

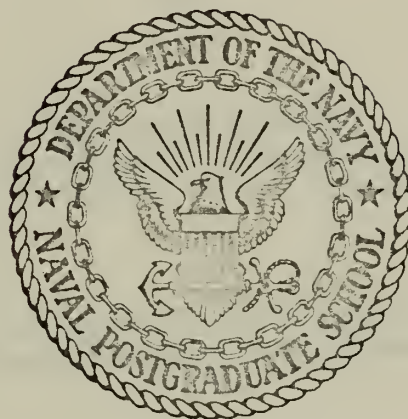
A HYBRID COMPUTER TECHNIQUE FOR  
MEASURING HUMAN DESCRIBING FUNCTIONS AND  
REMNANT IN CLOSED-LOOP TRACKING TASKS

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# NAVAL POSTGRADUATE SCHOOL

## Monterey, California



# THESIS

A HYBRID COMPUTER TECHNIQUE  
FOR  
MEASURING HUMAN DESCRIBING FUNCTIONS AND  
REMNANT IN CLOSED-LOOP TRACKING TASKS

by

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June 1972

T140015

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Measuring Human Describing Functions and  
Remnant in Closed-Loop Tracking Tasks

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Submitted in partial fulfillment of the  
requirements for the degree of

AERONAUTICAL ENGINEER.

from the

NAVAL POSTGRADUATE SCHOOL  
June 1972



## ABSTRACT

The measurement of the human describing function and remnant in a compensatory tracking task is undertaken.. These measurements are obtained through the application of the fast Fourier transform technique on a hybrid (analog-digital) computer. This method processes the data in real time with minimal core storage and the results are available immediately upon completion of the tracking run..





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# TABLE OF SYMBOLS

$A_{f_k}$	REAL PART OF FOURIER TRANSFORM OF $f(t)$ AT FREQUENCY $\omega_k$
$B_{f_k}$	IMAGINARY PART OF FOURIER TRANSFORM OF $f(t)$ AT FREQUENCY, $\omega_k$
$c(t)$	OUTPUT SIGNAL
$e(t)$	ERROR SIGNAL
$f(t)$	ARBITRARY SIGNAL
$F(j\omega)$	FOURIER TRANSFORM OF $f(t)$
$\bar{F}(j\omega)$	COMPLEX CONJUGATE OF $F(j\omega)$
$F(n)$	FOURIER COEFFICIENT FOR A PERIODIC SIGNAL $f(t)$
$\bar{F}(n)$	COMPLEX CONJUGATE OF $F(n)$
$H(s)$	SYSTEM TRANSFER FUNCTION
$i(t)$	INPUT SIGNAL
$i_T(t)$	INPUT SIGNAL OF FINITE DURATION
$n(t)$	REMNANT SIGNAL
$p(t)$	TOTAL OPERATOR RESPONSE SIGNAL, $p'(t) + n(t)$
$p'(t)$	LINEAR OPERATOR RESPONSE SIGNAL
$T$	PERIOD OF TOTAL RUN LENGTH
$T_k$	PERIOD FOR FREQUENCY, $\omega_k$
$Y_C(j\omega)$	CONTROLLED ELEMENT TRANSFER FUNCTION
$Y_P(j\omega)$	HUMAN DESCRIBING FUNCTION
$\phi_{ff}(\tau)$	AUTOCORRELATION FUNCTION OF $f(t)$
$\phi_{f_1 f_2}(\tau)$	CROSSCORRELATION FUNCTION OF $f_1(t)$ and $f_2(t)$
$\phi_{ff}(n)$	POWER SPECTRAL DENSITY OF PERIODIC SIGNAL, $f(t)$





- $\Phi_{f_1 f_2}^{(n)}$  CROSS-POWER SPECTRAL DENSITY OF PERIODIC SIGNALS  
 $f_1(t)$  and  $f_2(t)$
- $\Phi_{ff}(\omega)$  POWER (ENERGY) SPECTRAL DENSITY OF  $f(t)$
- $\Phi_{f_1 f_2}^{(\omega)}$  CROSS-POWER (ENERGY) SPECTRAL DENSITY OF  $f_1(t)$  and  
 $f_2(t)$
- $\omega_h$  FREQUENCIES OTHER THAN THOSE IN THE INPUT SINUSOIDS
- $\omega_k$  FREQUENCIES OF INPUT SINUSOIDS



## ACKNOWLEDGEMENTS

This thesis was written with the close support and never ending assistance of Asst. Professor R. A. Hess. The contributions of Asst. Professor M. H. Redlin as a second reader on this thesis are hereby acknowledged. In addition, the contributions of Mr. Robert L. Limes and Mr. Albert Wong of the Computer Laboratory staff are acknowledged for their assistance in the computer programming.



## I. INTRODUCTION

### A. BACKGROUND

Many of the important tasks performed by pilots are akin to those performed by linear servomechanisms. In situations such as this, the pilot can be modeled by a set of constant-coefficient linear differential equations. In the frequency domain, such a model is often referred to as a "human pilot describing function." The term "describing function" is preferred to "transfer function" to emphasize the fact that this pilot model is approximating a nonlinear element and is valid only for the particular inputs, system dynamics and task at hand.

The pilot-describing function is useful in studying two classes of problems. First, the describing functions measured in the piloted simulation of a given aircraft and task can be utilized in the subsequent stability and control analysis of this aircraft. Once the pilot's describing function for a particular task has been measured, he can be analytically replaced by his describing function in the analyses normally associated with the study of linear feedback systems. Second, actual measurement of pilot describing functions in ground simulation or in flight tests can be used to determine how a particular aircraft or flight task affects pilot behavior. Knowledge of pilot describing functions consequently provides valuable information for



the aircraft designer.

It is the problem of human describing function measurement which forms the basis of this research.. The hybrid computer program which has resulted can be utilized in virtually any study involving human behavior in compensatory tracking tasks.

#### B. COMPENSATORY TASKS AND QUASI-LINEARIZATION

The compensatory tracking task, shown in Figure 1,, assumes that the error signal is the only information that the operator is receiving. In this study the operator attempts to minimize a visual error signal by using a hand-operated controller. Tracking situations such as this are often encountered in aircraft flight control; e.g., a pilot attempting to maintain some desired pitch attitude in the presence of atmospheric turbulence.. It has been shown that in tasks such as this, the operator is nonlinear and time variant in behavior. He may, however, be successfully modeled in a quasi-linear fashion [Ref. 1].. This quasi-linearization implies that his response to visual stimulation is largely linear and time invariant; i.e., his dynamics are largely those of constant-coefficient linear differential equations. To account for nonlinear and/or time-varying behavior, the model includes a remnant signal as shown in Figure 2. The remnant is that portion of the operator's output which is not linearly correlated with the input. The human operator model thus consists of a linear describing function,  $Y_p(j\omega)$ , determined from the quasi-linear





analysis, and a remnant,  $n(t)$ .

It should be noted that this quasi-linear model is of little use if the remnant is relatively large, since then the operator's behavior is predominantly nonlinear and/or time varying.

To determine the human operator model it is necessary to calculate the linear describing function  $Y_p(j\omega)$ , and the remnant power spectral density,  $\Phi_{nn}(\omega)$ , from physical measurements of signals of finite duration in a laboratory experiment. The input signal must appear to the operator to be random, although it need not be truly random, and the operator must be well trained; i.e., not undergoing adaptation or learning [Ref. 2].

In order to measure the human describing function and remnant, one of three techniques can be employed. These are the direct Fourier analysis of the system signals, the use of crosscorrelation methods, and a model optimization technique. These methods are discussed in Ref. 3.

If, as done here, direct Fourier analysis or cross-correlation methods are employed in measuring the human describing function and remnant, then the concept of spectral analysis must be introduced. The next section is devoted to this topic. A more thorough treatment can be found in Ref. 4, Appendix D.



## II. SPECTRAL ANALYSIS

### A. PERIODIC SIGNALS

A periodic signal,  $f(t)$ , with a fundamental frequency  $\omega_1$  and period  $T$ , satisfying the Dirichlet conditions [Ref. 4, p. 579], may be represented by a Fourier series

$$f(t) = \sum_{n=-\infty}^{\infty} F(n) e^{jn\omega_1 t} \quad ,$$

where

$$F(n) = \frac{1}{T} \int_{-T/2}^{T/2} f(t) e^{-jn\omega_1 t} dt.$$

The autocorrelation function for the above periodic signal is defined as

$$\phi_{ff}(\tau) = \frac{1}{T} \int_{-T/2}^{T/2} f(t) f(t+\tau) dt.$$

This can be written

$$\phi_{ff}(\tau) = \frac{1}{T} \int_{-T/2}^{T/2} f(t) \sum_{n=-\infty}^{\infty} F(n) e^{jn\omega_1 (t+\tau)} dt$$

and is equivalent to

$$\phi_{ff}(\tau) = \left[ \sum_{n=-\infty}^{\infty} F(n) e^{jn\omega_1 \tau} \left( \frac{1}{T} \int_{-T/2}^{T/2} f(t) e^{jn\omega_1 t} dt \right) \right] .$$

With  $\bar{F}(n)$  denoting the complex conjugate of  $F(n)$ ,

$$\phi_{ff}(\tau) = \sum_{n=-\infty}^{\infty} F(n) e^{jn\omega_1 \tau} \bar{F}(n) = \sum_{n=-\infty}^{\infty} F(n) \bar{F}(n) e^{jn\omega_1 \tau} ,$$

$$\phi_{ff}(\tau) = \sum_{n=-\infty}^{\infty} |F(n)|^2 e^{jn\omega_1 \tau} .$$



The power spectral density,  $\Phi_{ff}(n)$ , is defined

$$\Phi_{ff}(n) = \frac{1}{T} \int_{-T/2}^{T/2} \phi_{ff}(\tau) e^{jn\omega_1 \tau} d\tau$$

and it can be shown that

$$\Phi_{ff}(n) = |F(n)|^2 \quad ..$$

It can be seen from this relationship that

$$\phi_{ff}(\tau) = \sum_{n=-\infty}^{\infty} \Phi_{ff}(n) e^{jn\omega_1 \tau} \quad ..$$

The crosscorrelation function,  $\phi_{f_1 f_2}(\tau)$ , of two periodic signals may be found in a similar manner if both signals have equal fundamental frequencies,  $\omega_1$ , and both signals satisfy the Dirichlet conditions. Assuming these conditions are met, then

$$\phi_{f_1 f_2}(\tau) = \frac{1}{T} \int_{-T/2}^{T/2} f_1(t) f_2(t+\tau) dt \quad ..$$

In a fashion similar to that for single signals, it can be shown

$$\phi_{f_1 f_2}(\tau) = \sum_{n=-\infty}^{\infty} \bar{F}_1(n) F_2(n) e^{jn\omega_1 \tau} \quad ..$$

The cross-power spectral density,  $\Phi_{f_1 f_2}(n)$ , is defined

$$\Phi_{f_1 f_2}(n) = \frac{1}{T} \int_{-T/2}^{T/2} \phi_{f_1 f_2}(\tau) e^{jn\omega_1 \tau} d\tau$$

and it can be shown that

$$\Phi_{f_1 f_2}(n) = \bar{F}_1(n) F_2(n) \quad ..$$

Using this relationship

$$\phi_{f_1 f_2}(\tau) = \sum_{n=-\infty}^{\infty} \Phi_{f_1 f_2}(n) e^{jn\omega_1 \tau} \quad .$$



## B. TRANSIENT SIGNALS

A signal,  $f(t)$ , is defined to be transient if

$$\lim_{t \rightarrow \infty} f(t) = 0.$$

If the transient signal,  $f(t)$ , satisfies the Dirichlet conditions in any finite interval, and if

$$\int_{-\infty}^{\infty} |f(t)| dt < \infty,$$

then the signal may be expressed as a Fourier integral [Ref.5, p. 279]. Under these conditions, the Fourier integral,

$$f(t) = \frac{1}{2\pi} \int_{-\infty}^{\infty} F(j\omega) e^{j\omega t} d\omega$$

gives the values of  $f(t)$  at all points, including those where the function is not continuous. The Fourier transform of  $f(t)$  is

$$F(j\omega) = \int_{-\infty}^{\infty} f(t) e^{-j\omega t} dt \quad ..$$

The autocorrelation function for the nonperiodic signal is defined

$$\phi_{ff}(\tau) = \int_{-\infty}^{\infty} f(t) f(t+\tau) dt \quad ..$$

This can also be written

$$\phi_{ff}(\tau) = \frac{1}{2\pi} \int_{-\infty}^{\infty} |F(j\omega)|^2 e^{j\omega \tau} d\omega \quad ..$$

Letting

$$\Phi_{ff}(\omega) = |F(j\omega)|^2 \quad ..$$

where  $\Phi_{ff}(\omega)$  is defined to be the energy spectral density of the signal  $f(t)$ , it can be shown that





$$\Phi_{ff}(\omega) = \int_{-\infty}^{\infty} \phi_{ff}(\tau) e^{-j\omega\tau} d\tau \quad ..$$

Thus it can be seen that the energy density spectrum and the autocorrelation function of a transient signal are a Fourier transform pair,

$$\phi_{ff}(\tau) = \frac{1}{2\pi} \int_{-\infty}^{\infty} \Phi_{ff}(\omega) e^{j\omega\tau} d\omega$$

and

$$\Phi_{ff}(\omega) = \int_{-\infty}^{\infty} \phi_{ff}(\tau) e^{-j\omega\tau} d\tau \quad ..$$

If two transient signals,  $f_1(t)$  and  $f_2(t)$ , each satisfy the Dirichlet conditions in all finite intervals, and if

$$\int_{-\infty}^{\infty} |f_1(t)| dt < \infty \quad \text{and} \quad \int_{-\infty}^{\infty} |f_2(t)| dt < \infty \quad ,$$

then

$$\phi_{f_1 f_2}(\tau) = \int_{-\infty}^{\infty} f_1(t) f_2(t+\tau) dt = \frac{1}{2\pi} \int_{-\infty}^{\infty} \bar{F}_1(j\omega) F_2(j\omega) e^{j\omega\tau} d\omega \quad ..$$

Now with

$$\Phi_{f_1 f_2}(\omega) = \bar{F}_1(j\omega) F_2(j\omega) \quad ,$$

where  $\Phi_{f_1 f_2}(\omega)$  is defined as the cross-energy spectral density of the signals  $f_1(t)$  and  $f_2(t)$ , it can be shown that

$$\Phi_{f_1 f_2}(\omega) = \int_{-\infty}^{\infty} \phi_{f_1 f_2}(\tau) e^{j\omega\tau} d\tau \quad ,$$

and has as its inverse transform

$$\phi_{f_1 f_2}(\tau) = \frac{1}{2\pi} \int_{-\infty}^{\infty} \Phi_{f_1 f_2}(\omega) e^{j\omega\tau} d\omega \quad ..$$

The transient signals  $f_1(t)$  and  $f_2(t)$  are said to be linearly uncorrelated when  $\phi_{f_1 f_2}(\tau) = 0$  for all  $\tau$ .



### C. RANDOM SIGNALS

In general, a random signal,  $f(t)$ , from a stationary, ergodic random process [Ref. 4, p 279], does not have a Fourier transform since

$$\int_{-\infty}^{\infty} |f(t)| dt$$

is not finite. An autocorrelation function may be defined for the random signal  $f(t)$  as

$$\phi_{ff}(\tau) = \lim_{T \rightarrow \infty} \frac{1}{2T} \int_{-T}^T f(t) f(t+\tau) dt \quad ..$$

Since  $\phi_{ff}(\tau)$  satisfies the Dirichlet conditions for all finite intervals and

$$\int_{-\infty}^{\infty} |\phi_{ff}(\tau)| d\tau < \infty \quad ..$$

it can be represented by a Fourier integral. It can then be shown that

$$\phi_{ff}(\tau) = \frac{1}{2\pi} \int_{-\infty}^{\infty} \Phi_{ff}(\omega) e^{j\omega\tau} d\omega \quad ..$$

where  $\Phi_{ff}(\omega)$  is defined as the power spectral density of the signal  $f(t)$ .  $\Phi_{ff}(\omega)$  is the Fourier transform of  $\phi_{ff}(\tau)$ ,

$$\Phi_{ff}(\omega) = \int_{-\infty}^{\infty} \phi_{ff}(\tau) e^{-j\omega\tau} d\tau \quad ..$$

If there exist two random signals,  $f_1(t)$  and  $f_2(t)$ , which are sample functions from two different random processes, each of which are stationary, ergodic, and jointly ergodic, then the crosscorrelation function is given as

$$\phi_{f_1 f_2}(\tau) = \lim_{T \rightarrow \infty} \frac{1}{2T} \int_{-T}^T f_1(t) f_2(t+\tau) dt = \frac{1}{2\pi} \int_{-\infty}^{\infty} \Phi_{f_1 f_2}(\omega) e^{j\omega\tau} d\omega, ..$$



where  $\phi_{f_1 f_2}(\omega)$  is defined as the cross-power spectral density of the random signals  $f_1(t)$  and  $f_2(t)$ ..  $\phi_{f_1 f_2}(\tau)$  has the Fourier transform

$$\phi_{f_1 f_2}(\omega) = \int_{-\infty}^{\infty} \phi_{f_1 f_2}(\tau) e^{-j\omega\tau} d\tau \quad ..$$

The random signals  $f_1(t)$  and  $f_2(t)$  are said to be linearly uncorrelated if  $\phi_{f_1 f_2}(\tau) = 0$  for all  $\tau$ ..



### III. DESCRIBING FUNCTION AND REMNANT RELATIONS

#### A. FREQUENCY DOMAIN RELATIONS

Again consider Figure 2 with the input a sample function from an ergodic random process. It is seen that

$$E(j\omega) = I(j\omega) - C(j\omega) ,$$

where the Fourier transforms are as defined in Section B. Now

$$C(j\omega) = \left[ N(j\omega) + Y_p(j\omega)E(j\omega) \right] Y_c(j\omega) ,$$

then

$$E(j\omega) = \frac{\left[ I(j\omega) - N(j\omega)Y_c(j\omega) \right]}{\left[ 1 + Y_p(j\omega)Y_c(j\omega) \right]} ..$$

Finally, after multiplying by  $\bar{I}(j\omega)$ , the complex conjugate of the input,

$$\bar{I}(j\omega)E(j\omega) = \frac{\bar{I}(j\omega) \left[ I(j\omega) - N(j\omega)Y_c(j\omega) \right]}{\left[ 1 + Y_p(j\omega)Y_c(j\omega) \right]} ..$$

In a similar manner,

$$P(j\omega) = Y_p(j\omega)E(j\omega) + N(j\omega)$$

and then substituting for  $E(j\omega)$  from above,

$$P(j\omega) = \frac{\left[ Y_p(j\omega)I(j\omega) - N(j\omega)Y_p(j\omega)Y_c(j\omega) \right]}{\left[ 1 + Y_p(j\omega)Y_c(j\omega) \right]} + N(j\omega)$$

and

$$\bar{I}(j\omega)P(j\omega) = \frac{\bar{I}(j\omega) \left[ Y_p(j\omega)I(j\omega) + N(j\omega) \right]}{\left[ 1 + Y_p(j\omega)Y_c(j\omega) \right]} ..$$

Likewise,  $\bar{I}(j\omega)C(j\omega)$ ,  $\bar{E}(j\omega)P(j\omega)$ ,  $\bar{E}(j\omega)E(j\omega)$ ,  $\bar{P}(j\omega)P(j\omega)$ , and  $\bar{C}(j\omega)C(j\omega)$  may be calculated. The results, to be utilized shortly, are shown in Table I.





## B. FINITE RUN LENGTH

Since it is impossible to have experimental runs of infinite duration, measurements using finite time histories are necessary. Reference 6 indicates that finite run lengths can be handled analytically as follows: If  $i_T(t)$  is the input and defined

$$i_T(t) = \begin{cases} i(t) & , -T \leq t \leq T \\ 0 & , \text{ ELSEWHERE} \end{cases}$$

then this function can be considered to be transient and have a Fourier transform

$$I(j\omega) = \int_{-\infty}^{\infty} i_T(t) e^{-j\omega t} dt \quad ..$$

The other system signals and their transforms can be defined in precisely the same manner. If the run time,  $T$ , is large enough to ensure accurate power spectral measurements, yet finite so that the respective Fourier transforms exist, then the following spectral relations are valid [Ref.. 6].

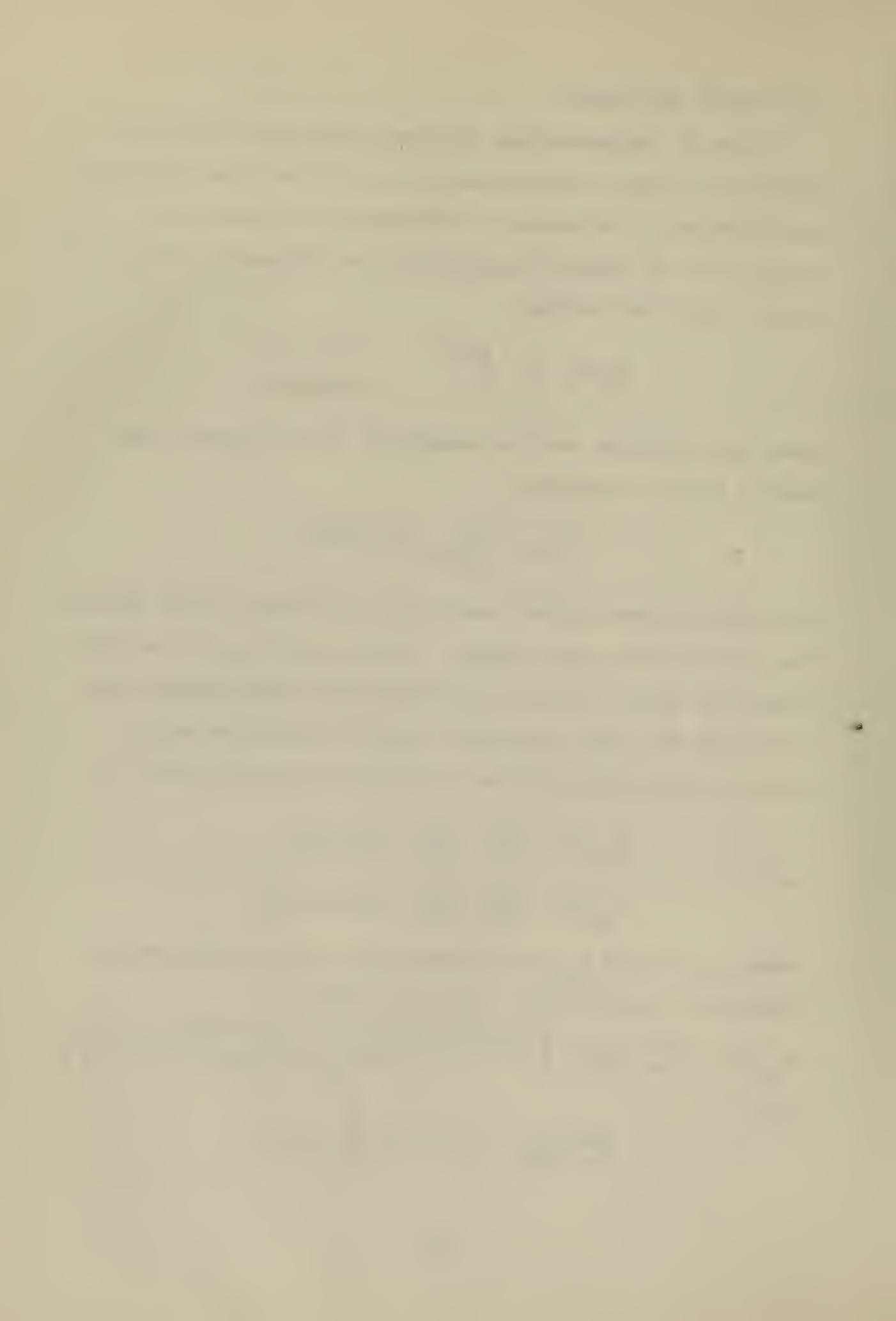
$$\begin{aligned} \Phi_{ii}(\omega) &= \lim_{T \rightarrow \infty} \left[ \frac{1}{2T} |I(j\omega)|^2 \right] \\ \text{and} \\ \Phi_{ip}(\omega) &= \lim_{T \rightarrow \infty} \left[ \frac{1}{2T} \bar{I}(j\omega) P(j\omega) \right] \end{aligned}$$

where  $\Phi_{ii}(\omega)$  and  $\Phi_{ip}(\omega)$  are power and cross-power spectral densities respectively. Utilizing Table I,

$$\Phi_{ip}(\omega) = \lim_{T \rightarrow \infty} \left\{ \frac{1}{2T} \left[ \frac{\bar{I}(j\omega) I(j\omega) Y_p(j\omega) + \bar{I}(j\omega) N(j\omega) Y_c(j\omega)}{\bar{I} + Y_p(j\omega) Y_c(j\omega)} \right] \right\}$$

but

$$\lim_{T \rightarrow \infty} \left[ \frac{1}{2T} \bar{I}(j\omega) I(j\omega) \right] = \Phi_{ii}(\omega)$$



and

$$\lim_{T \rightarrow \infty} \left[ \frac{1}{2T} \bar{I}(j\omega) N(j\omega) \right] = \Phi_{in}(\omega)$$

thus

$$\Phi_{ip}(\omega) = \frac{Y_p(j\omega) \Phi_{ii}(\omega) + \Phi_{in}(\omega) Y_c(j\omega)}{1 + Y_p(j\omega) Y_c(j\omega)} \quad ..$$

Now,

$$\Phi_{in}(\omega) = \int_{-\infty}^{\infty} \phi_{in}(\omega) e^{-j\omega\tau} d\tau \quad ..$$

Since by definition the remnant,  $n(t)$ , is linearly uncorrelated with the input,  $i(t)$ , then  $\phi_{in}(\tau) = 0$  for all  $\tau$ . Thus

$$\Phi_{in}(\omega) = 0 \text{ and}$$

$$\Phi_{ip}(\omega) = \frac{Y_p(j\omega) \Phi_{ii}(\omega)}{1 + Y_p(j\omega) Y_c(j\omega)}$$

In a like manner,

$$\Phi_{ie}(\omega) = \lim_{T \rightarrow \infty} \left[ \frac{1}{2T} \bar{I}(j\omega) E(j\omega) \right] \quad ..$$

or

$$\Phi_{ie}(\omega) = \frac{\Phi_{ii}(\omega) - \Phi_{in}(\omega) Y_c(j\omega)}{1 + Y_p(j\omega) Y_c(j\omega)} \quad ..$$

Again since  $n(t)$  and  $i(t)$  are linearly uncorrelated,

$$\phi_{in}(\tau) = 0 \text{ for all } \tau, \text{ then } \Phi_{in}(\omega) = 0. \text{ Thus}$$

$$\Phi_{ie}(\omega) = \frac{\Phi_{ii}(\omega)}{1 + Y_p(j\omega) Y_c(j\omega)} \quad ..$$

Also

$$\Phi_{pp}(\omega) = \lim_{T \rightarrow \infty} \left[ \frac{1}{2T} \bar{P}(j\omega) P(j\omega) \right]$$

$$\Phi_{pp}(\omega) = \lim_{T \rightarrow \infty} \left\{ \frac{1}{2T} \left[ \frac{\bar{I}(j\omega) \bar{Y}_p(j\omega) + \bar{N}(j\omega)}{1 + \bar{Y}_p(j\omega) \bar{Y}_c(j\omega)} \right] \left[ \frac{I(j\omega) Y_p(j\omega) + N(j\omega)}{1 + Y_p(j\omega) Y_c(j\omega)} \right] \right\},$$



or

$$\Phi_{pp}(\omega) = \frac{\Phi_{ii}(\omega) |Y_p(j\omega)|^2 + \Phi_{ni}(\omega) Y_p(j\omega) + \Phi_{in}(\omega) \bar{Y}_p(j\omega) + \Phi_{nn}(\omega)}{|1 + Y_p(j\omega) Y_c(j\omega)|^2} \quad ..$$

Again, since  $\Phi_{in}(\omega) = \Phi_{ni}(\omega) = 0$ ,

$$\Phi_{pp}(\omega) = \frac{\Phi_{ii}(\omega) |Y_p(j\omega)|^2 + \Phi_{nn}(\omega)}{|1 + Y_p(j\omega) Y_c(j\omega)|^2}$$

### C. DESCRIBING FUNCTION AND REMNANT

From  $\Phi_{ip}(\omega)$ ,  $\Phi_{ie}(\omega)$ , and  $\Phi_{pp}(\omega)$ ,  $Y_p(j\omega)$  and  $\Phi_{nn}(\omega)$  may be found. Utilizing  $\Phi_{ie}(\omega)$  and  $\Phi_{ip}(\omega)$ ,

$$\Phi_{ii}(\omega) = \Phi_{ie}(\omega) |1 + Y_p(j\omega) Y_c(j\omega)|$$

and

$$\Phi_{ii}(\omega) = \frac{1}{Y_p(j\omega)} \Phi_{ip}(\omega) |1 + Y_p(j\omega) Y_c(j\omega)| \quad ..$$

Thus,

$$\Phi_{ie}(\omega) |1 + Y_p(j\omega) Y_c(j\omega)| = \frac{1}{Y_p(j\omega)} \Phi_{ip}(\omega) |1 + Y_p(j\omega) Y_c(j\omega)| \quad ..$$

or

$$Y_p(j\omega) = \frac{\Phi_{ip}(\omega)}{\Phi_{ie}(\omega)} \quad ..$$

Also, from  $\Phi_{pp}(\omega)$ ,

$$\Phi_{nn}(\omega) = |1 + Y_p(j\omega) Y_c(j\omega)|^2 \Phi_{pp}(\omega) - |Y_p(j\omega)|^2 \Phi_{ii}(\omega) \quad ..$$

In addition,  $Y_c(j\omega)$  can be determined and calculated from

$$\Phi_{ic}(\omega) = \Phi_{in}(\omega) Y_c(j\omega) + \Phi_{ie}(\omega) Y_p(j\omega) Y_c(j\omega) \quad ..$$

Again  $\Phi_{in}(\omega) = 0$ , thus



$$\Phi_{ic}(\omega) = \Phi_{ie}(\omega) Y_p(j\omega) Y_c(j\omega) \quad ..$$

But

$$\Phi_{ie}(\omega) Y_p(j\omega) = \Phi_{ip}(\omega) \quad ,$$

thus

$$Y_c(j\omega) = \frac{\Phi_{ic}(\omega)}{\Phi_{ip}(\omega)} \quad ..$$

The functions in Table II form the basis of the describing function measurement technique utilized in this study.

#### D. SINUSOIDAL INPUTS

The relations in Table II are predicated on the existence of a random input. In experimental work, a random appearing input is often used and can be generated as a summation of sine waves [Ref. 3]. The input can thus be represented by

$$i(t) = \sum_{k=1}^n A_k \sin \omega_k t$$

where the  $\omega_k$  are chosen such that in a finite run length there will exist an integral number of periods or complete cycles. In this analysis a run time of 150 seconds was used and  $0.08 \leq \omega_k \leq 40.0$  radians per second.

Utilization of a sinusoidal input results in system signals that have both random and periodic components. With what is now a mixed signal, the question arises as to which power spectral relationship should be used. The solution is to use the periodic power spectral relationships for measurements made at the input frequencies and to use the random, finite relationships at all other frequencies.





It should be noted that if the experimental run length,  $T$ , is large and contains an integral number of periods of each of the input sinusoids, then the Fourier transforms of the periodic and finite random signals differ only by a constant of proportionality [Ref. 6]..

In the periodic case, it was shown that

$$F_k(n) = \frac{1}{2T_k} \int_{-T_k}^{T_k} f(t) e^{-jn\omega_k t} dt, ,$$

and for the random signal,

$$F(j\omega) = \int_{-\infty}^{\infty} f_T(t) e^{-j\omega t} dt \quad ..$$

If it is recalled that  $f_T(t) = 0$  for  $t < -T$  and  $t > T$ , and letting  $T = m_k T_k$ , where  $m_k$  is the number of periods,  $T_k$ , of frequency  $\omega_k$ , then

$$F_k(n) = \frac{1}{2T_k} \int_{-T_k}^{T_k} f(t) e^{-jn\omega_k t} dt = \frac{1}{2m_k T_k} \int_{-m_k T_k}^{m_k T_k} f(t) e^{-jn\omega_k t} dt, ,$$

or

$$F_k(n) = \frac{1}{2T} \int_{-T}^T f(t) e^{-j\omega_k t} dt \quad ; ;$$

thus it can be seen that

$$2TF_k(n) = F(j\omega_k) \quad ..$$

It should be noted that the same expansion of the limits on the integral can be used for the periodic function; thus

$$\Phi_{ff}(n_k) = \frac{1}{2T} \int_{-T}^T \frac{1}{2T} \int_{-T}^T f(t) f(t+\tau) dt e^{-jn\omega_k \tau} d\tau, ,$$



or

$$\Phi_{ff}(n_k) = \frac{1}{4T^2} \int_{-T}^T \int_{-T}^T f(t) f(t+\tau) e^{-jn_k \tau} dt d\tau \quad ..$$

Also for the finite random function,

$$\Phi_{ff}(\omega)_T = \int_{-T}^T \phi_{ff}(\tau)_T e^{-j\omega \tau} d\tau \quad ..$$

or

$$\Phi_{ff}(\omega)_T = \frac{1}{2T} \int_{-T}^T \int_{-T}^T f_T(t) f_T(t+\tau) e^{-j\omega \tau} dt d\tau \quad ;;$$

then by equating the integrals, since  $f_T(t) = f(t)$  for  $-T \leq t \leq T$ , it is seen that at a specific input frequency,  $\omega_k$ ,

$$\Phi_{ff}(\omega_k)_T = 2T \Phi_{ff}(n_k) \quad ..$$

where

$$\Phi_{ff}(\omega_k)_T = \lim_{T \rightarrow \infty} \left[ \frac{1}{2T} |F(j\omega_k)|^2 \right] \quad ..$$

It has been shown that

$$Y_p(j\omega) = \frac{\Phi_{ip}(\omega)}{\Phi_{ie}(\omega)} \quad ..$$

At the input frequencies, with  $T$  large and containing an integral number of periods of each input frequency,

$$Y_p(j\omega_k) = \frac{\Phi_{ip}(\omega_k)}{\Phi_{ie}(\omega_k)} = \frac{\Phi_{ip}(\omega_k)_T}{\Phi_{ie}(\omega_k)_T} = \frac{\Phi_{ip}(n_k) 2T}{\Phi_{ie}(n_k) 2T} \quad .$$

or

$$Y_p(j\omega_k) = \frac{\Phi_{ip}(n_k)}{\Phi_{ie}(n_k)} = \frac{\bar{I}(n_k) P(n_k)}{\bar{I}(n_k) E(n_k)} = \frac{P(n_k)}{E(n_k)} \quad ..$$

Similarly,



$$Y_C(j\omega_k) \doteq \frac{\Phi_{ic}(n_k)}{\Phi_{ip}(n_k)} = \frac{C(n_k)}{P(n_k)} \quad .$$

This illustrates that the cross-power spectral measurements need not be made and only the Fourier transforms are needed. The latter is usually an easier measurement than the former.

The terms in  $\Phi_{nn}(\omega)$  can be examined in a similar manner. In this case, if measurements are taken at frequencies,  $\omega_h$ , other than those used in the input, then the expression for  $\Phi_{nn}(\omega)$  is somewhat simplified; i.e.,

$$\Phi_{nn}(\omega_h) = \left| 1 + Y_p(j\omega_h) Y_C(j\omega_h) \right|^2 \Phi_{pp}(\omega_h) \quad ,$$

since  $\Phi_{ii}(\omega_h) = 0$ .

It should be emphasized that at frequencies,  $\omega_h$ , other than those used in the input,

$$Y_p(j\omega_h) \neq \frac{P(n_h)}{E(n_h)} \quad \text{and} \quad Y_C(j\omega_h) \neq \frac{C(n_h)}{P(n_h)} \quad ,$$

since  $I(n_h) = \bar{I}(n_h) = 0$ . Thus  $Y_p(j\omega_h)$  and  $Y_C(j\omega_h)$  must be estimated, as direct calculations can be made only at the input frequencies. In order to estimate  $Y_p(j\omega_h)$ , simple linear interpolation between  $Y_p(j\omega_k)$  and  $Y_p(j\omega_{k+1})$  can be used.



#### IV. COMPUTER MECHANIZATION

##### A. EXPERIMENTAL SET-UP

The measurement of a human's describing function and remnant in the compensatory tracking task of Figure 2 was made using a hybrid (analog-digital) computer. The error signal,  $e(t)$ , was viewed as the vertical displacement of a horizontal line on an oscilloscope screen. The operator's controller consisted of a non-moving force stick. Control was effected by fore and aft pressure on the stick; e.g., if the line on the oscilloscope moved above the datum, the operator applied forward pressure to move the line down, and vice-versa. The input,  $i(t)$ , and controlled element dynamics,  $Y_c(j\omega)$ , were mechanized on the computer as were the measurement algorithms to be described. Each experimental tracking run lasted 150 seconds.

##### B. FAST FOURIER TRANSFORM

The pertinent relationships are again

$$Y_p(j\omega_k) = \frac{P(n_k)}{E(n_k)}$$

and

$$\Phi_{nn}(\omega_h) = \left| 1 + Y_p(j\omega_h) Y_c(j\omega_h) \right|^2 \Phi_{pp}(\omega_h) \quad .$$

Now

$$P(n_k) = \frac{1}{T} \int_{-T/2}^{T/2} p(t) e^{-j\omega_k t} dt \quad ,$$





or

$$P(n_k) = \frac{1}{T} \left[ \int_{-T/2}^{T/2} p(t) \cos(n\omega_k t) dt + j \int_{-T/2}^{T/2} p(t) \sin(n\omega_k t) dt \right] \dots$$

These integrals may be approximated by the following summations:

$$P(n_k) \doteq \frac{\Delta t}{T} \sum_{n=0}^N p(n\Delta t) \cos(\omega_k n\Delta t) + j \frac{\Delta t}{T} \sum_{n=0}^N p(n\Delta t) \sin(\omega_k n\Delta t) \dots$$

If

$$A_{p_k} = \sum_{n=0}^N p(n\Delta t) \cos(\omega_k n\Delta t)$$

and

$$B_{p_k} = \sum_{n=0}^N p(n\Delta t) \sin(\omega_k n\Delta t) \dots$$

then the fast Fourier transform  $P(n_k)$  can be written as

$$P(n_k) \doteq \frac{\Delta t}{T} \left[ A_{p_k} + j B_{p_k} \right] \dots$$

Similarly, it can be shown that

$$C(n_k) \doteq \frac{\Delta t}{T} \left[ A_{c_k} + j B_{c_k} \right] \quad \text{and} \quad E(n_k) \doteq \frac{\Delta t}{T} \left[ A_{e_k} + j B_{e_k} \right] \dots$$

From this,

$$Y_p(j\omega_k) \doteq \frac{A_{p_k} + j B_{p_k}}{A_{e_k} + j B_{e_k}}$$

and

$$|Y_p(j\omega_k)| = \left[ \frac{A_{p_k}^2 + B_{p_k}^2}{A_{e_k}^2 + B_{e_k}^2} \right]^{1/2} \dots$$

It can also be shown that

$$\angle Y_p(j\omega_k) = \tan^{-1} \frac{B_{e_k}}{A_{e_k}} - \tan^{-1} \frac{B_{p_k}}{A_{p_k}} \dots$$



In order to validate the simulated controlled element dynamics,  $Y_c(j\omega)$ , on-line measurement of  $Y_c(j\omega)$  can be utilized:

$$|Y_c(j\omega_k)| = \left[ \frac{A_{c_k}^2 + B_{c_k}^2}{A_{p_k}^2 + B_{p_k}^2} \right]^{1/2}$$

and

$$\angle Y_c(j\omega_k) = \tan^{-1} \frac{B_{p_k}}{A_{p_k}} - \tan^{-1} \frac{B_{c_k}}{A_{c_k}} \quad ..$$

The power spectra of the remnant can also be determined from the above relationships with the interpolation process described earlier. The determination of the power spectra of the operator output,  $\Phi_{pp}(\omega_h)$ , can be accomplished using the measurements  $p(t)$  at any desired frequency [Ref. 6].. Previously it was shown that

$$\Phi_{pp}(\omega_h) \doteq \Phi_{pp}(n_h)T = \Phi_{pp}(n_h)2T \quad ..$$

thus

$$\Phi_{pp}(\omega_h) \doteq \bar{P}(n_h)P(n_h)2T = 2T|P(n_h)|^2 = \frac{(\Delta t)^2}{T^2} 2T|A_{p_h}^2 + B_{p_h}^2| \quad ,$$

and finally,

$$\Phi_{pp}(\omega_h) \doteq \frac{2(\Delta t)^2}{T} |A_{p_h}^2 + B_{p_h}^2| \quad ..$$

From this it can be seen that in order to determine the operator's describing function,  $Y_p(j\omega_k)$ , the remnant,  $\Phi_{nn}(\omega_h)$ , and the controlled element dynamics,  $Y_c(j\omega_k)$ , the only measurements needed are those of the error, operator output, and controlled element output. If each of those measurements, taken at specific times, is then multiplied



by the proper trigonometric function and summed over the entire run, then the describing functions and remnant can be calculated.

The major drawback of the fast Fourier transform technique is the necessity of calling the trigonometric functions  $\sin(x)$  and  $\cos(x)$  during the run.. This requires so much computer time that the multiplication and addition computations cannot be performed between analog-to-digital interrupts. This in turn means that all data must be stored for later computational purposes.

The necessity of calling trigonometric functions during the run can be avoided if the relationships for  $\sin(a+b)$  and  $\cos(a+b)$  are utilized in recursive fashion.. If the initial time of the run is "a" and the time between measurements is "b" then

$$\sin(a+b) = \sin(a)\cos(b) + \cos(a)\sin(b)$$

and

$$\cos(a+b) = \cos(a)\cos(b) - \sin(a)\sin(b) ..$$

It is obvious that a recursive process can be mechanized which obviates calling trigonometric functions during the run.

The use of this method yields results immediately upon completion of the run and conserves computer storage space. Within seconds of run completion, the data may be analyzed from either numerical or graphical read-out..



## V. RESULTS

The operation of the program was checked by measuring the transfer function of known elements or filters in place of the operator. The results are shown in Figure 3, 4, 5, and 6.

In validating the remnant measurement technique, the operator was again replaced by an element with known transfer function. A signal

$$n(t) = A \sin(\omega_h t), \quad \omega_k < \omega_h < \omega_{k+1} \quad "$$

was inserted at the output of this element. Measured values of  $\Phi_{nn}(\omega)$  were then compared with the theoretical  $\frac{A^2}{4}$  value [Ref. 7].

Actual describing function and remnant measurements were then taken on two subjects. The first of these had considerable previous experience. The second subject had experience only with this experimental set-up. The measurement results are shown in Figures 7-18 for controlled elements of  $Y_c(s) = 1.0$  and  $\frac{1}{s}$ . Each of these figures represents the results of 10 tracking runs. In all runs, the human describing function contains the gain of the controller.

The describing functions illustrated are comparable with those obtained by other experimenters; e.g., [Ref. 8].

This computer program represents a powerful tool for use in experimental investigations involving a human







controller. It can, for example, be utilized in a variety of situations in which quantitative models of pilot behavior are desired. The program itself requires little core storage.. This means that considerable storage is available for simulating complex aircraft dynamics, providing detailed display formats and calculating performance measures..



TABLE I

$$E(j\omega) = \frac{I(j\omega) - N(j\omega)Y_c(j\omega)}{1 + Y_p(j\omega)Y_c(j\omega)}$$

$$P(j\omega) = \frac{I(j\omega)Y_p(j\omega) + N(j\omega)}{1 + Y_p(j\omega)Y_c(j\omega)}$$

$$C(j\omega) = N(j\omega)Y_c(j\omega) + E(j\omega)Y_p(j\omega)Y_c(j\omega)$$

$$\bar{I}(j\omega)E(j\omega) = \frac{\bar{I}(j\omega)I(j\omega) - \bar{I}(j\omega)N(j\omega)Y_c(j\omega)}{1 + Y_p(j\omega)Y_c(j\omega)}$$

$$\bar{I}(j\omega)P(j\omega) = \frac{\bar{I}(j\omega)I(j\omega)Y_p(j\omega) + \bar{I}(j\omega)N(j\omega)}{1 + Y_p(j\omega)Y_c(j\omega)}$$

$$\bar{I}(j\omega)C(j\omega) = [\bar{I}(j\omega)N(j\omega) + \bar{I}(j\omega)E(j\omega)Y_p(j\omega)]Y_c(j\omega)$$

$$\bar{E}(j\omega)E(j\omega) = \left[ \frac{\bar{I}(j\omega) - \bar{N}(j\omega)\bar{Y}_c(j\omega)}{1 + \bar{Y}_p(j\omega)\bar{Y}_c(j\omega)} \right] \left[ \frac{I(j\omega) - N(j\omega)Y_c(j\omega)}{1 + Y_p(j\omega)Y_c(j\omega)} \right]$$

$$\bar{P}(j\omega)P(j\omega) = \left[ \frac{\bar{I}(j\omega)\bar{Y}_p(j\omega) + \bar{N}(j\omega)}{1 + \bar{Y}_p(j\omega)\bar{Y}_c(j\omega)} \right] \left[ \frac{I(j\omega)Y_p(j\omega) + N(j\omega)}{1 + Y_p(j\omega)Y_c(j\omega)} \right]$$

$$\bar{C}(j\omega)C(j\omega) = [\bar{N}(j\omega) + \bar{E}(j\omega)\bar{Y}_p(j\omega)][\bar{Y}_c(j\omega)][N(j\omega) + E(j\omega)Y_p(j\omega)][Y_c(j\omega)]$$



TABLE II

$$Y_p(j\omega) = \frac{\Phi_{ip}(\omega)}{\Phi_{ie}(\omega)}$$

$$Y_c(j\omega) = \frac{\Phi_{ic}(\omega)}{\Phi_{ip}(\omega)}$$

$$\Phi_{nn}(\omega) = \Phi_{pp}(\omega) |1 + Y_p(j\omega) Y_c(j\omega)|^2 - \Phi_{ii}(\omega) |Y_p(j\omega)|^2$$

$$\Phi_{pp}(\omega) = \frac{\Phi_{ii}(\omega) |Y_p(j\omega)|^2 + \Phi_{nn}(\omega)}{|1 + Y_p(j\omega) Y_c(j\omega)|^2}$$



# HUMAN OPERATOR in a COMPENSATORY TRACKING TASK

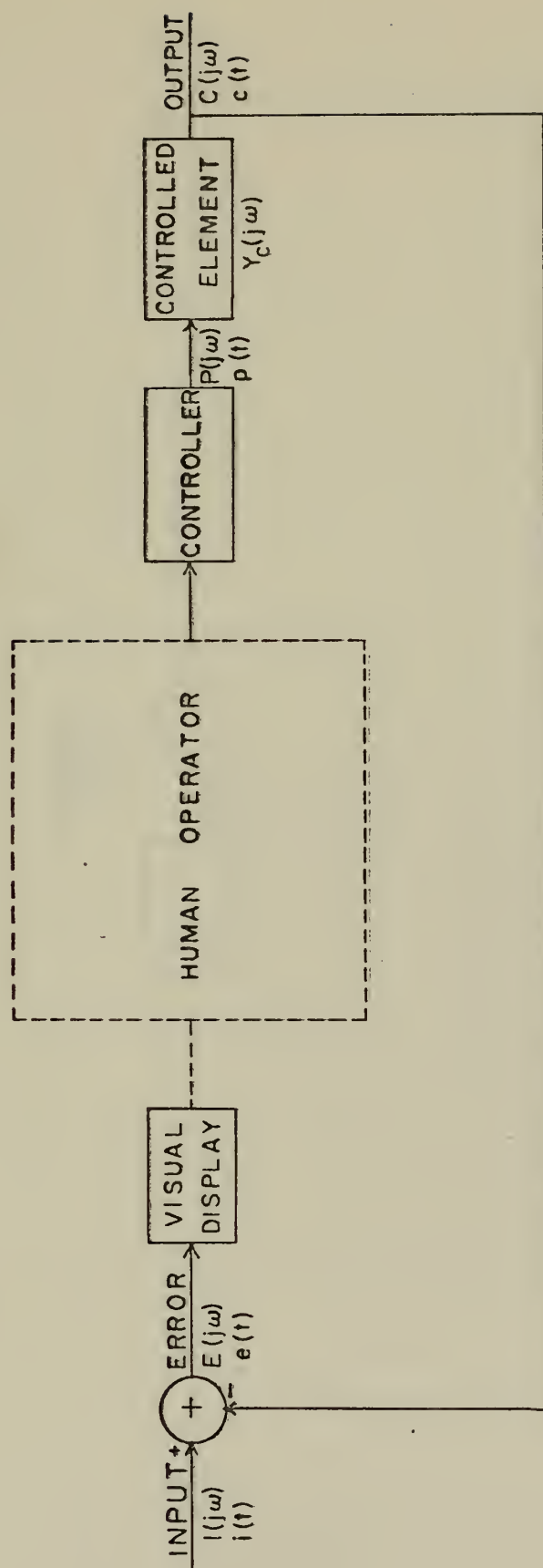


FIGURE 1





# QUASI-LINEAR MODEL of HUMAN OPERATOR in a COMPENSATORY TRACKING TASK

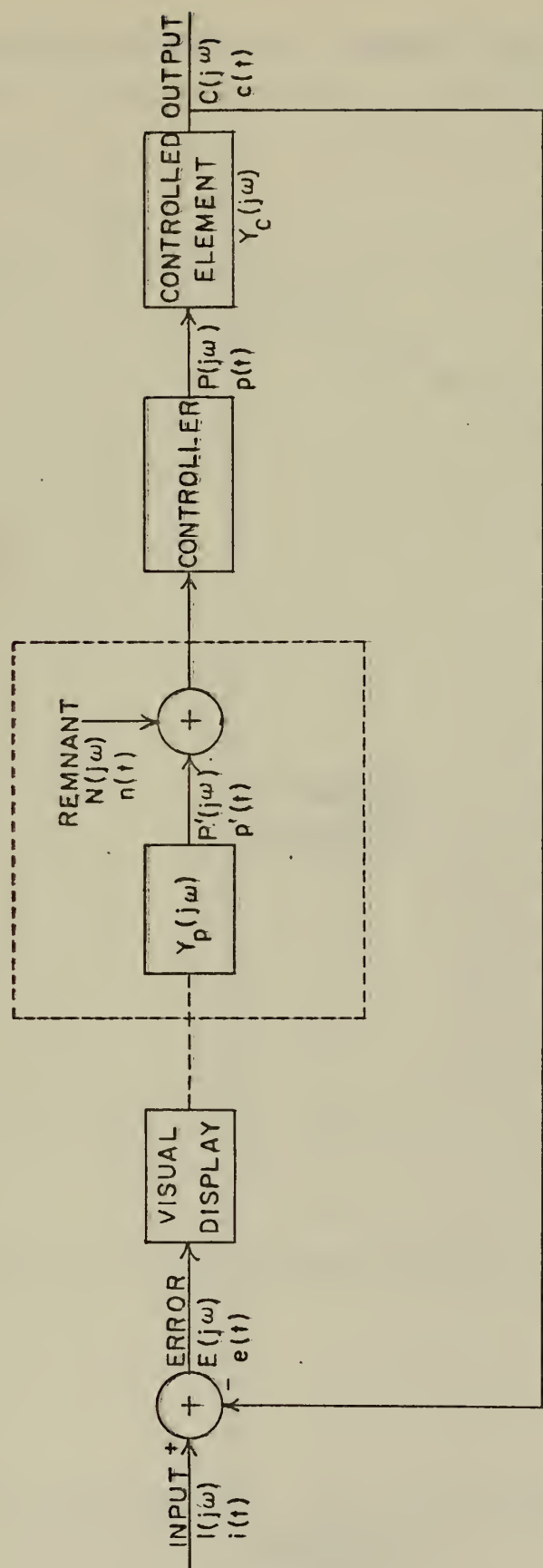


FIGURE 2



# MEASUREMENT of KNOWN ( $\frac{1}{s}$ ) ELEMENT

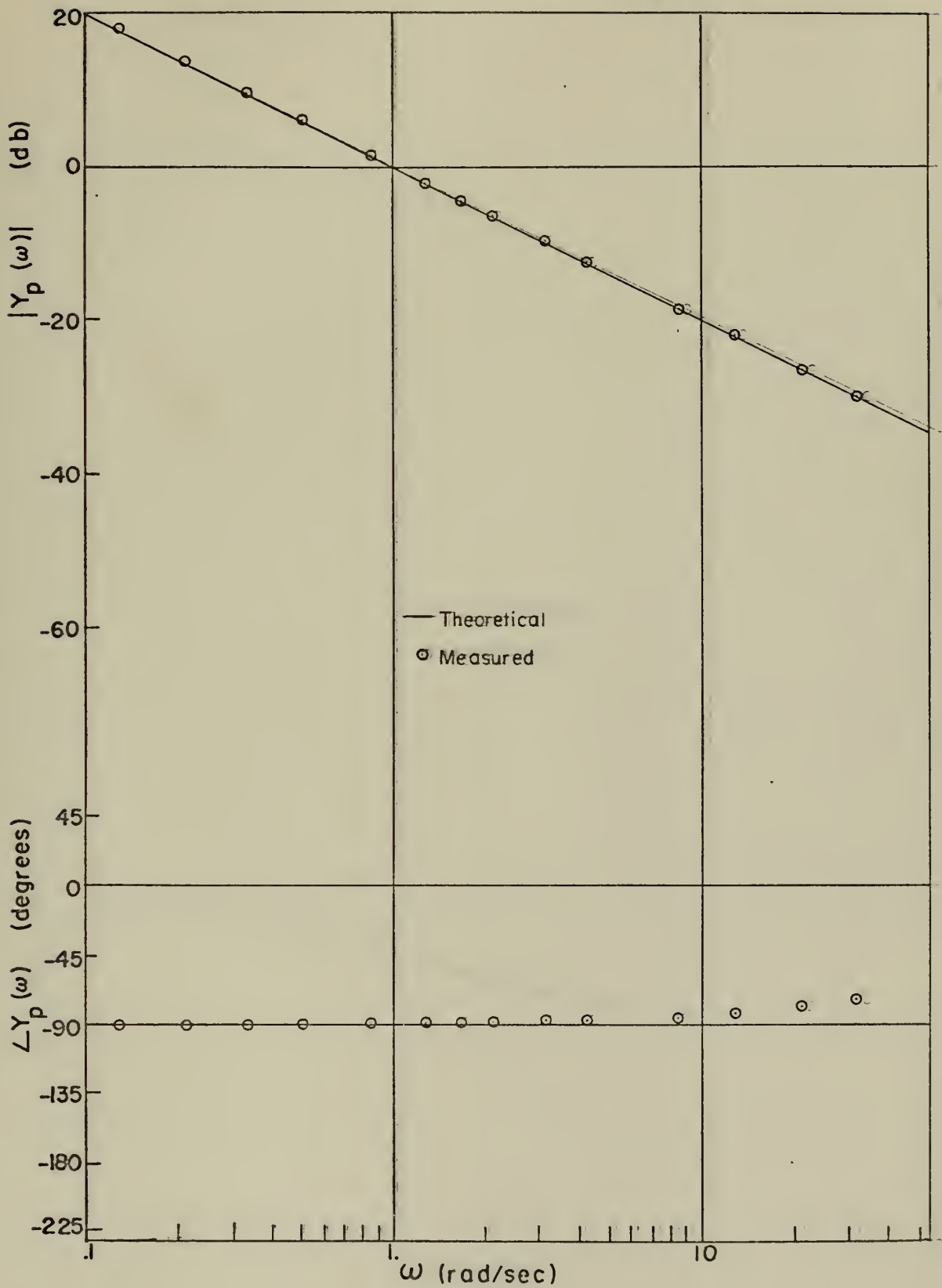


FIGURE 3



# MEASUREMENT of KNOWN $\left(\frac{1}{s+1}\right)$ ELEMENT

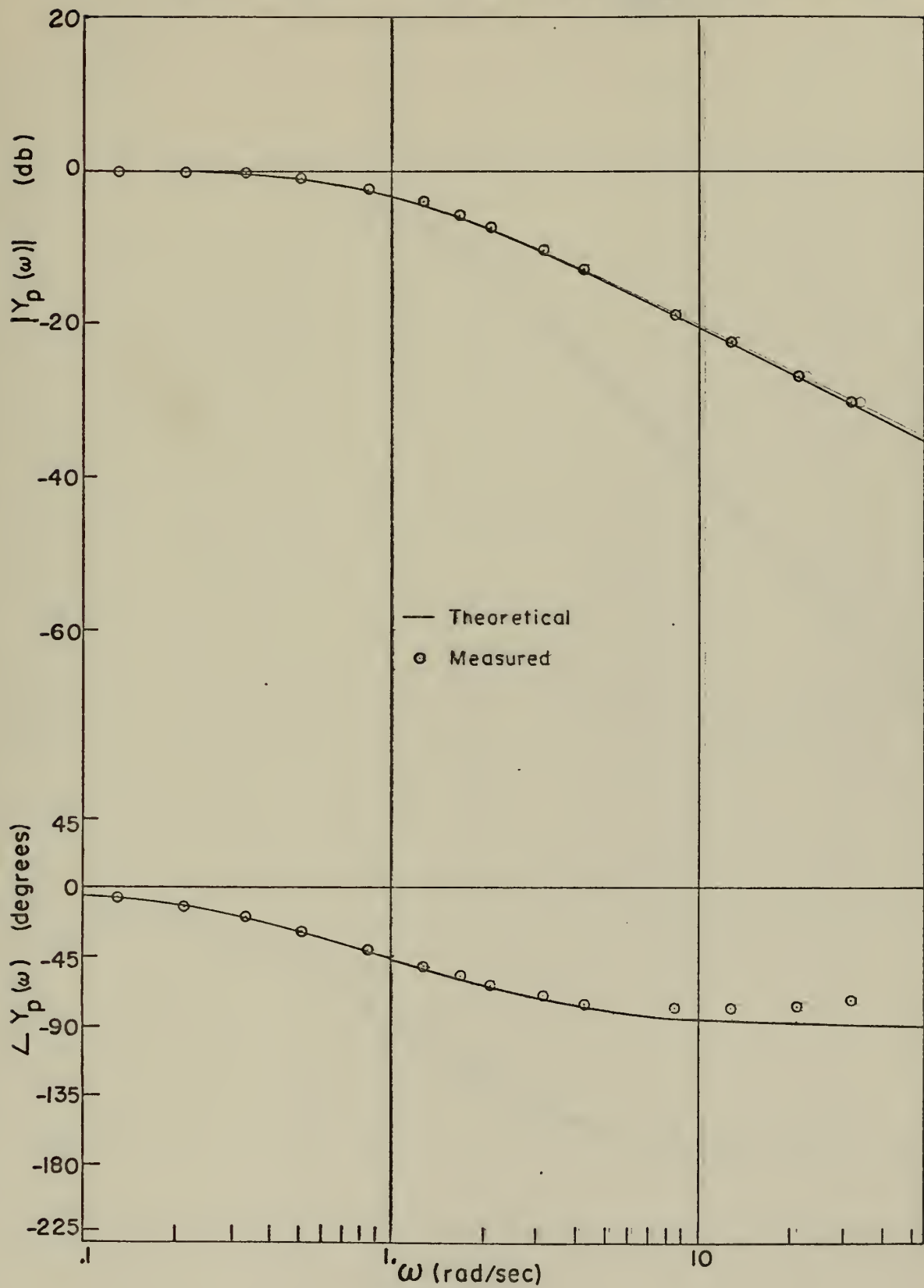


FIGURE 4



MEASUREMENT of KNOWN  $\left(\frac{1}{(s+1)^2}\right)$  ELEMENT

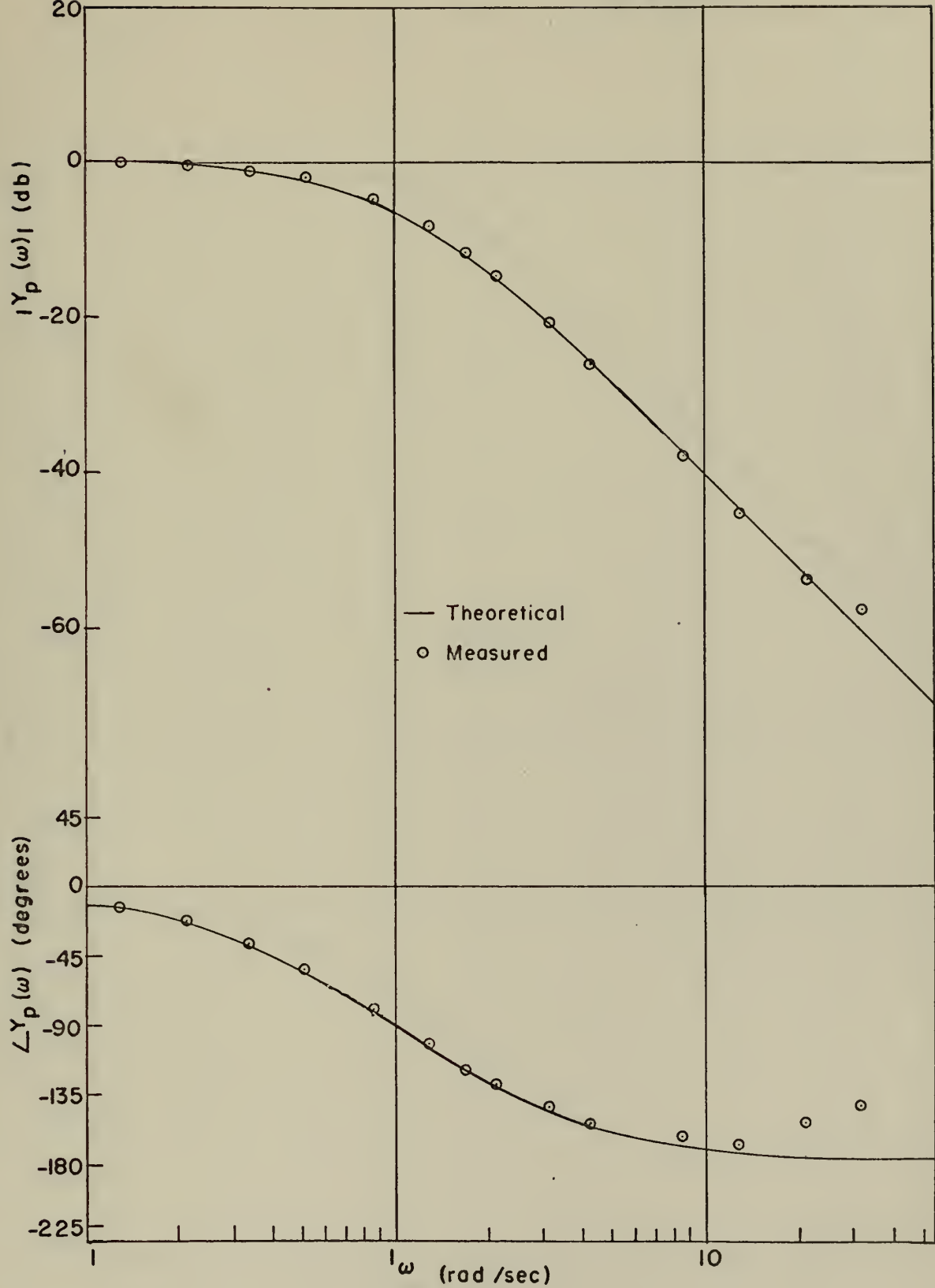


FIGURE 5





MEASUREMENT of KNOWN  $\left(\frac{2}{s^2 + 2s + 2}\right)$  ELEMENT

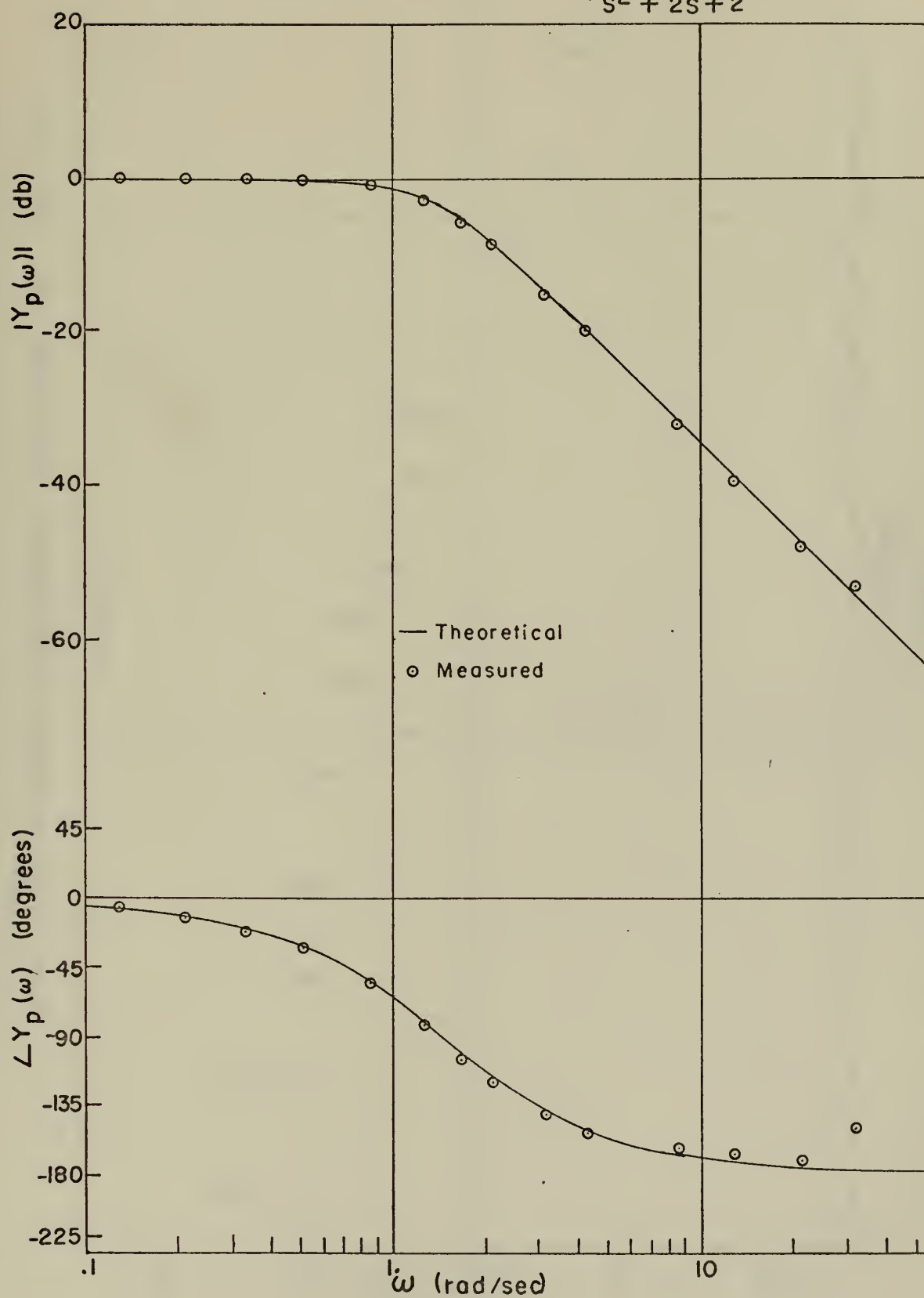
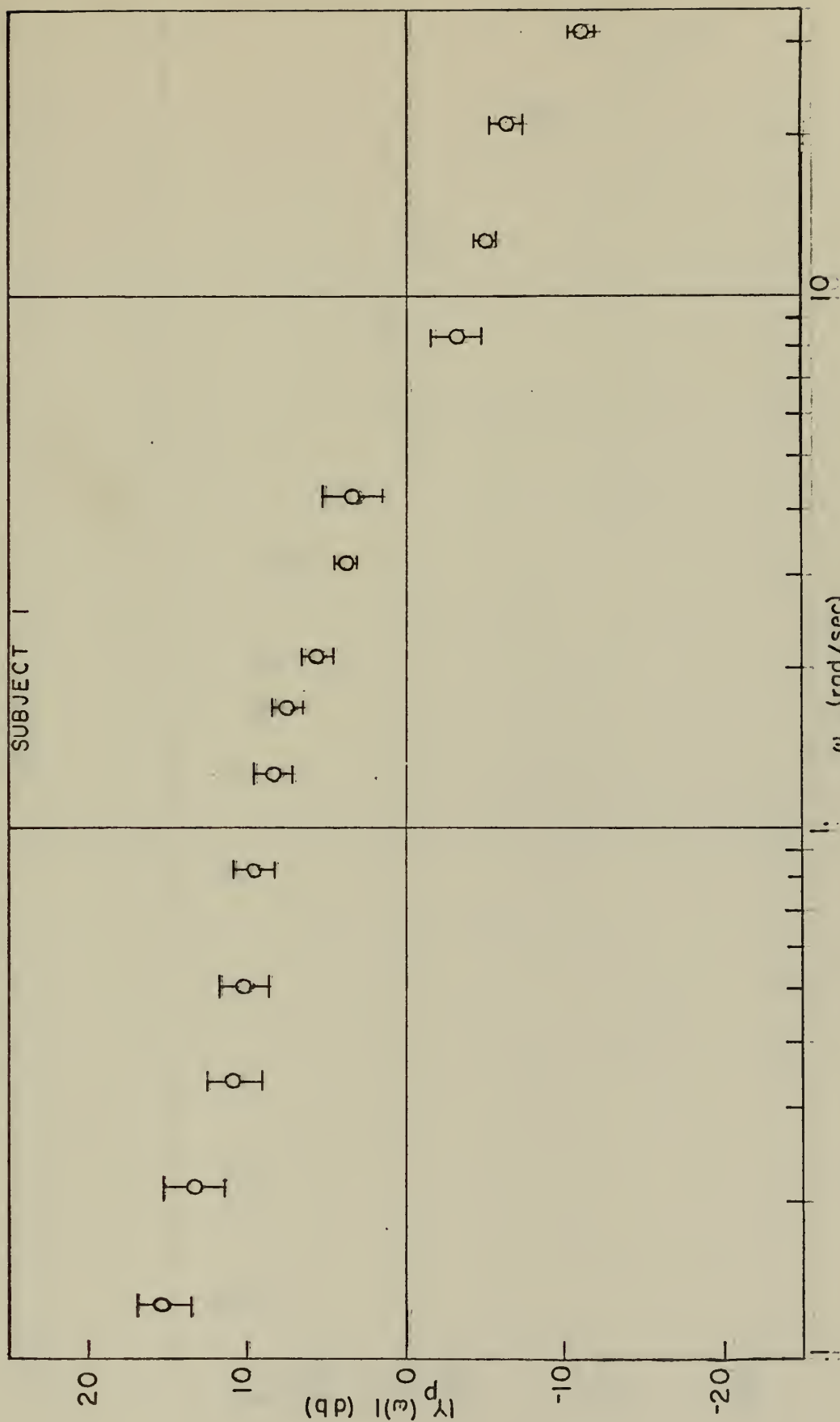


FIGURE 6



AMPLITUDE of HUMAN DESCRIBING FUNCTION for  $Y_C(s) = 1.0$



$\omega$  (rad/sec)  
FIGURE 7



PHASE of HUMAN DESCRIBING FUNCTION for  $Y_c(s) = 1.0$

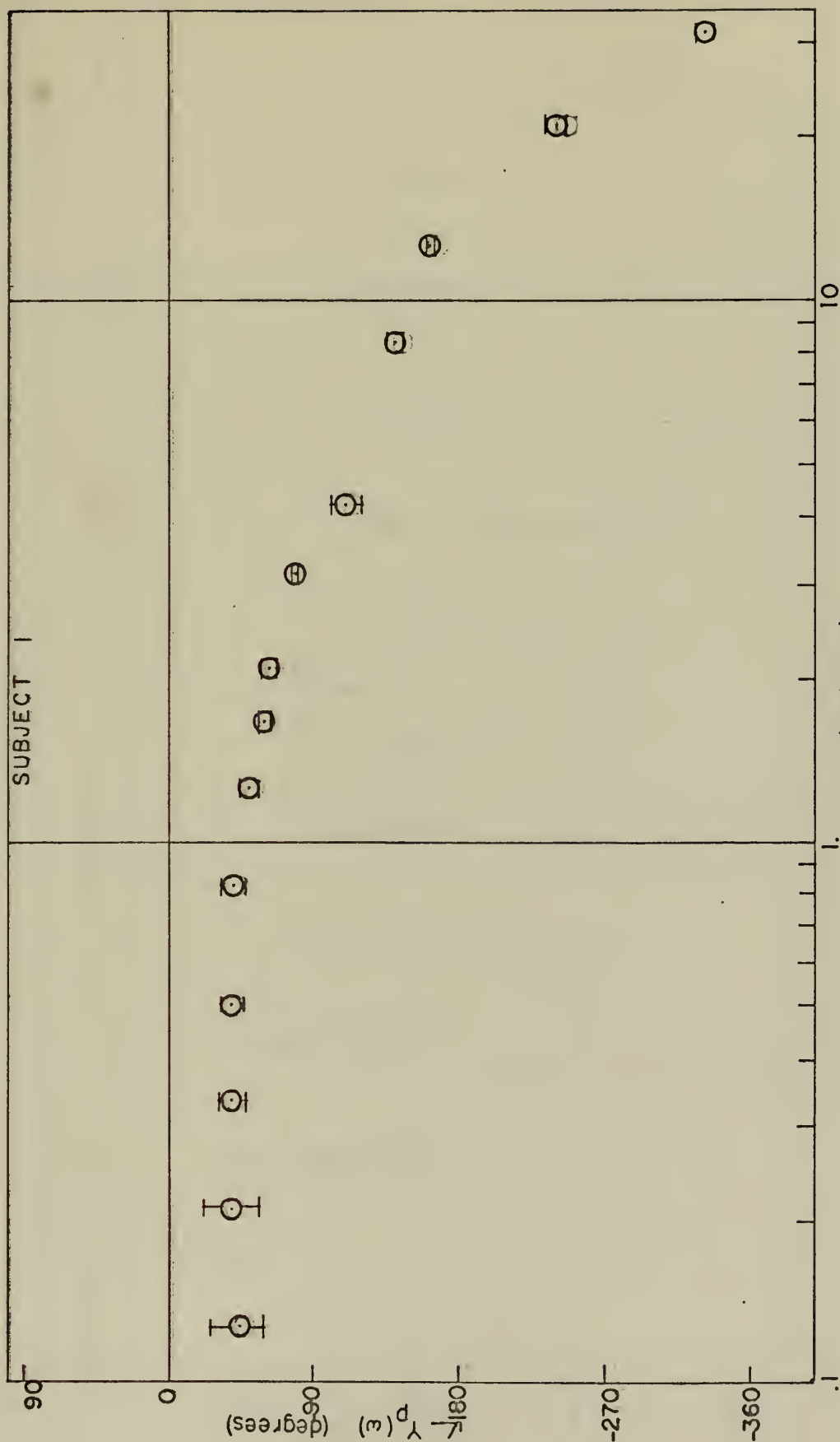
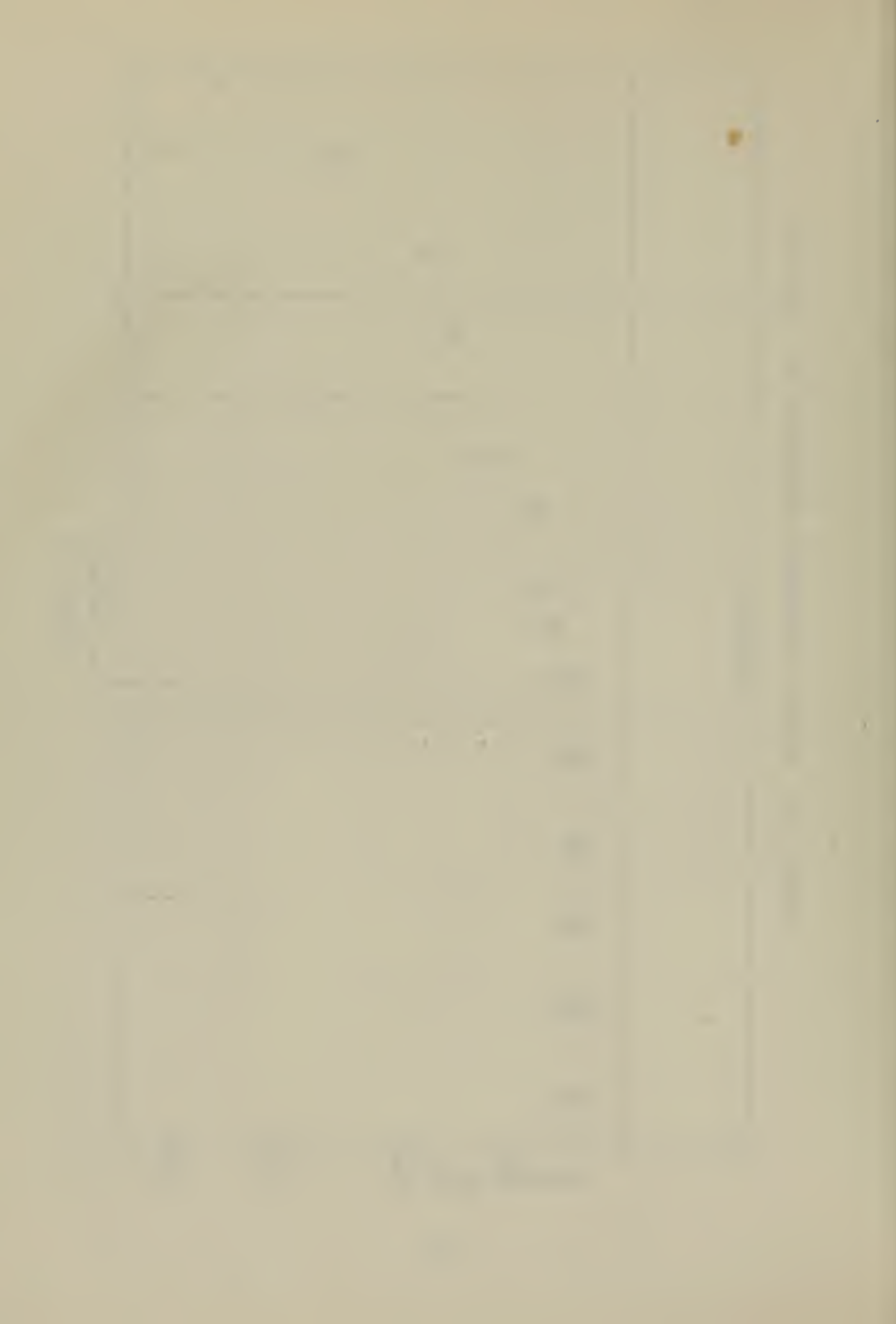
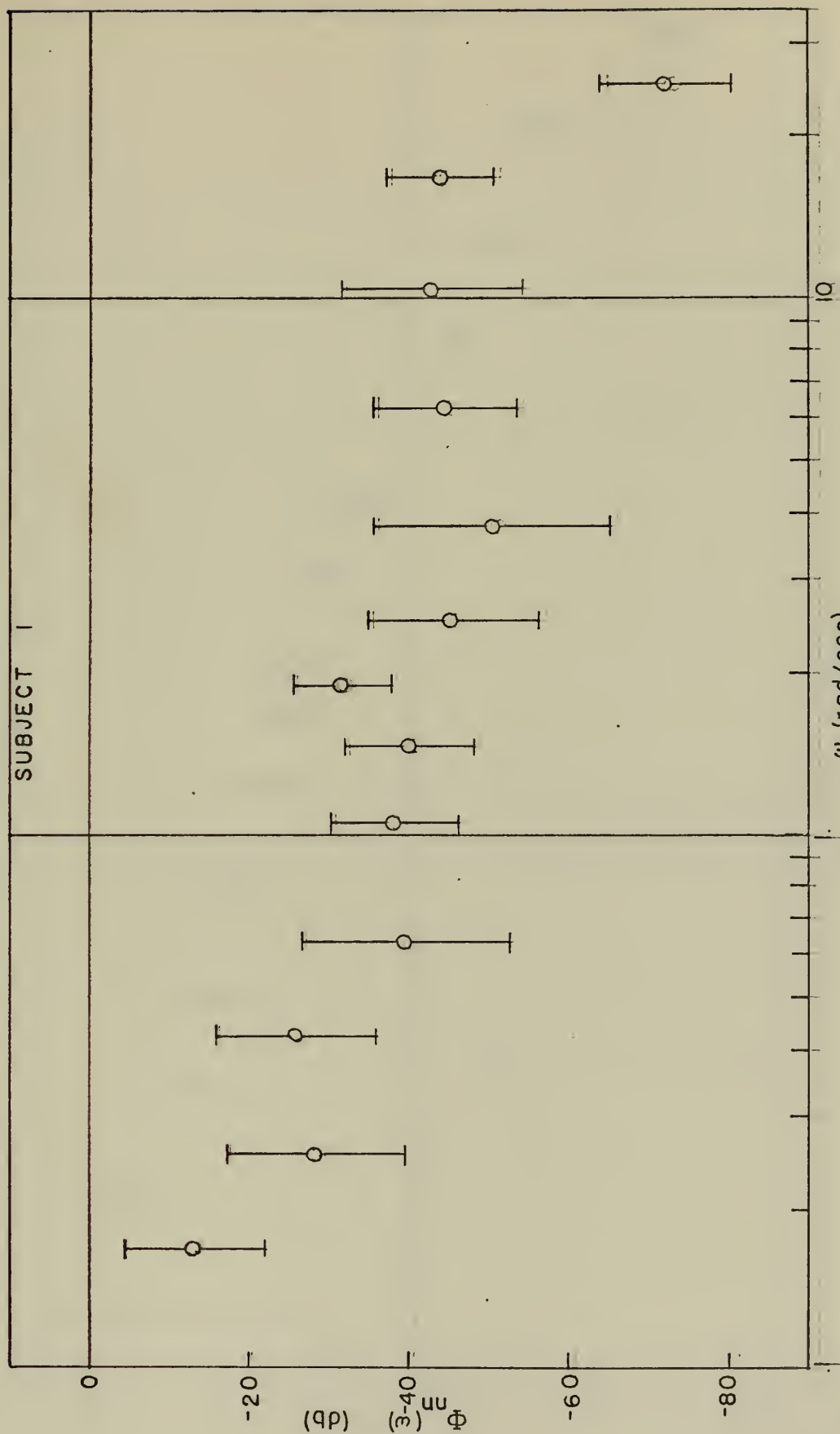


FIGURE 8



# REMNANT of HUMAN DESCRIBING FUNCTION for $Y_c(s) = 1.0$



$\omega$  (rad/sec)  
FIGURE 9





AMPLITUDE of HUMAN DESCRIBING FUNCTION for  $Y_c(s) = 1.0$

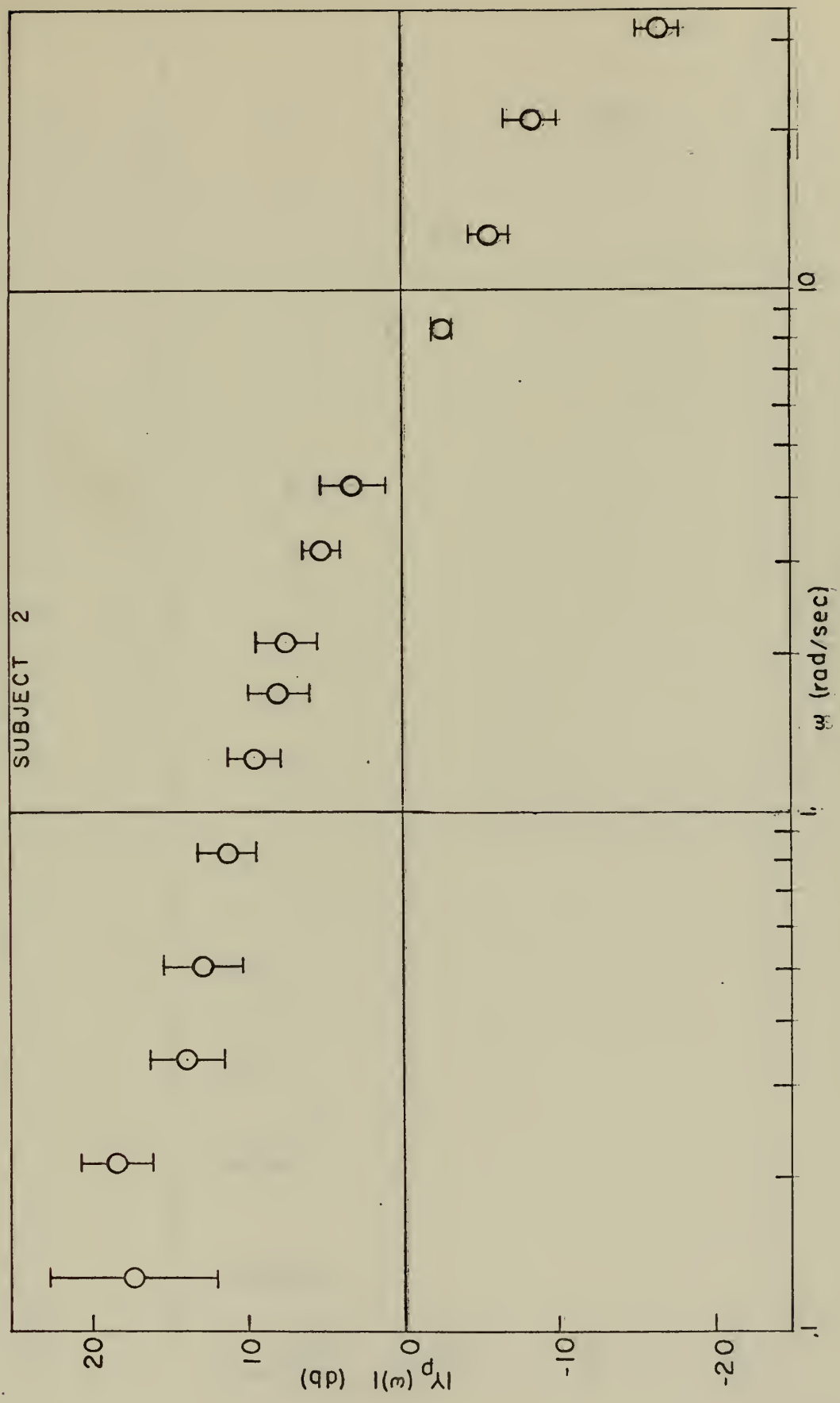
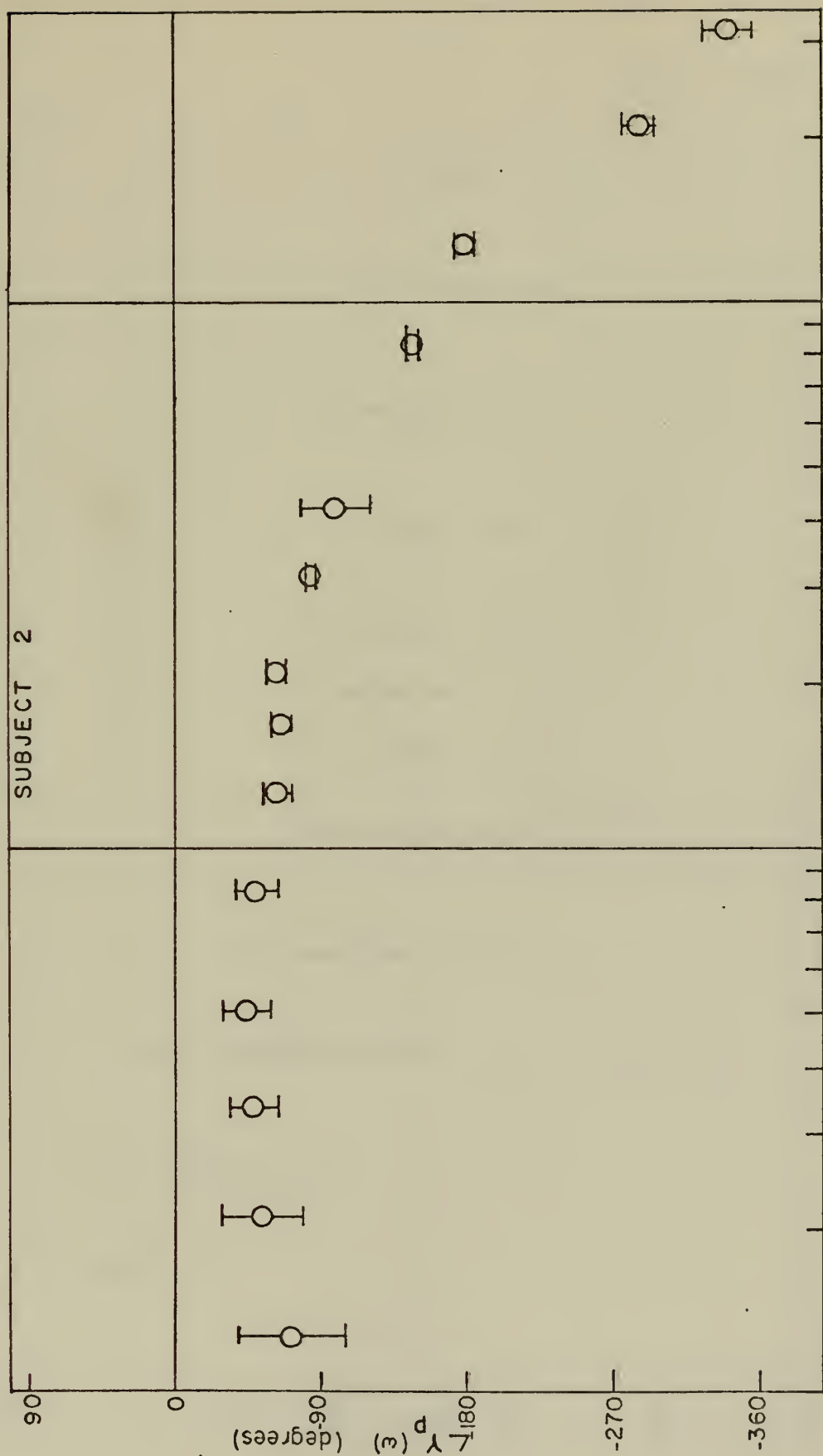


FIGURE 10



PHASE of HUMAN DESCRIBING FUNCTION for  $Y_c(s) = 1.0$



$\omega$  (rad/sec)  
FIGURE 11



# REMNANT of HUMAN DESCRIBING FUNCTION for $Y_C(s) = 1.0$

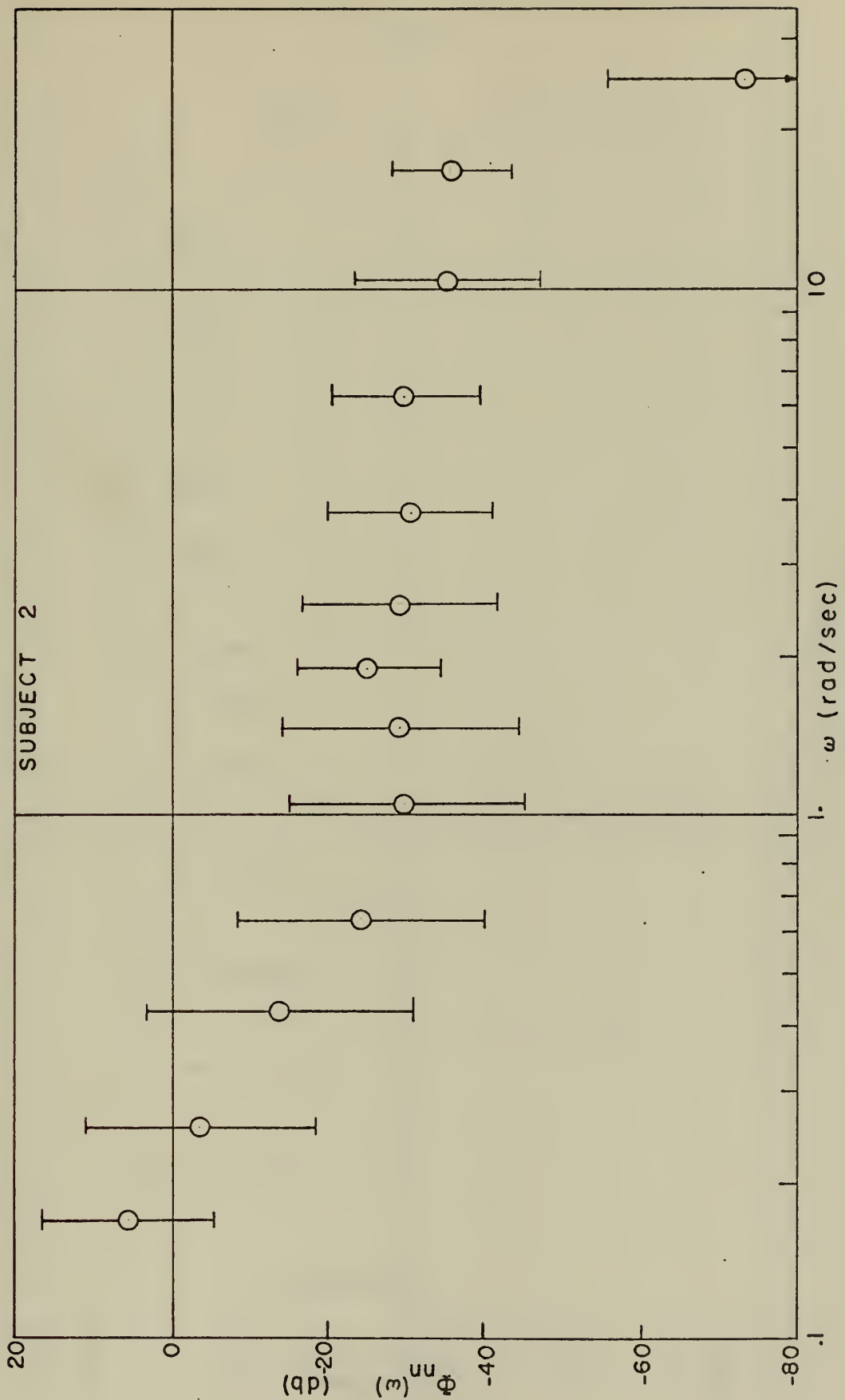


FIGURE 12



AMPLITUDE of HUMAN DESCRIBING FUNCTION for  $Y_c(s) = \frac{1}{s}$

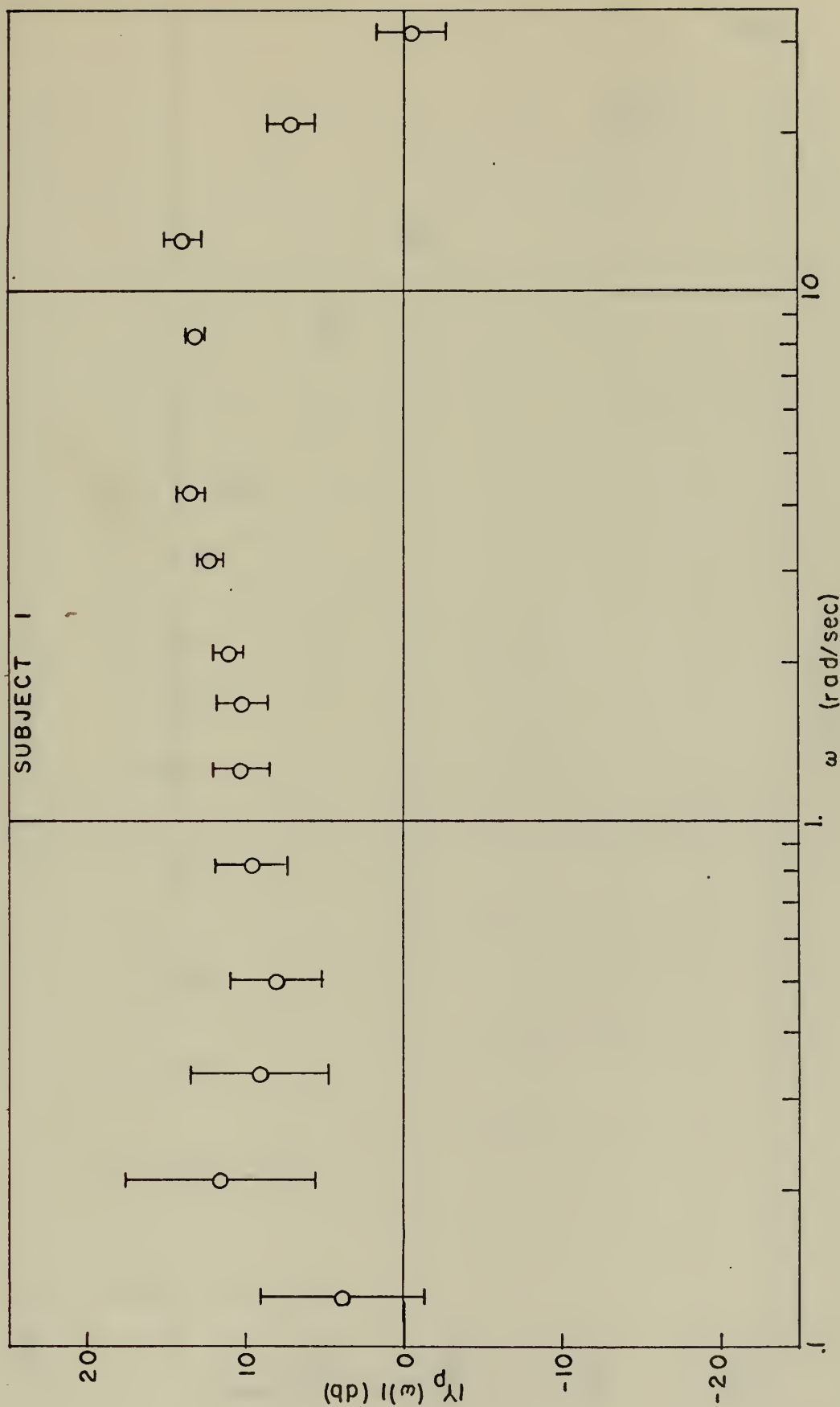
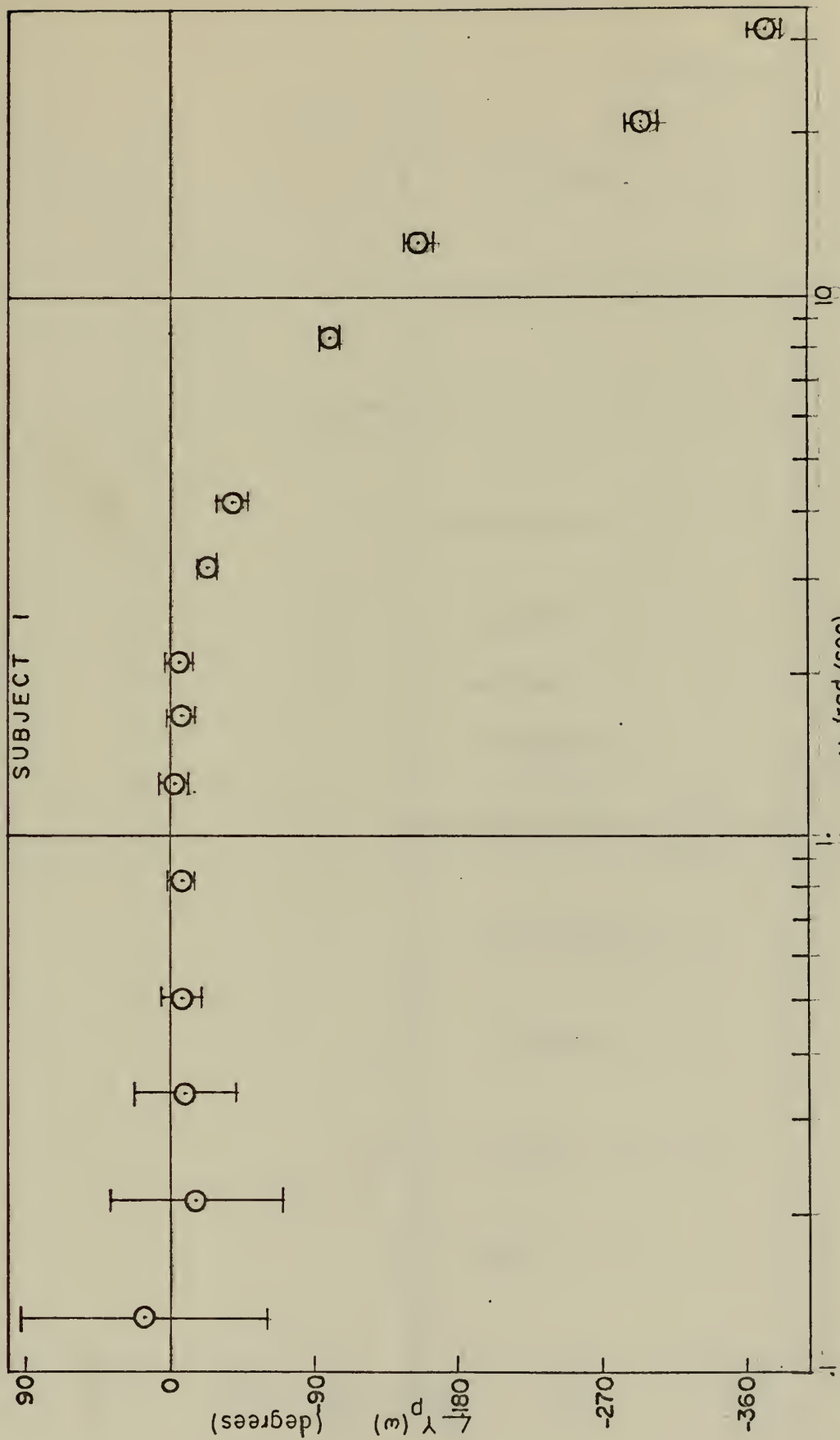


FIGURE 13





PHASE of HUMAN DESCRIBING FUNCTION for  $\gamma_c(s) = \frac{1}{s}$



$\omega$  (rad/sec)  
FIGURE 14



# REMNANT OF HUMAN DESCRIBING FUNCTION for $Y_c(s) = \frac{1}{s}$

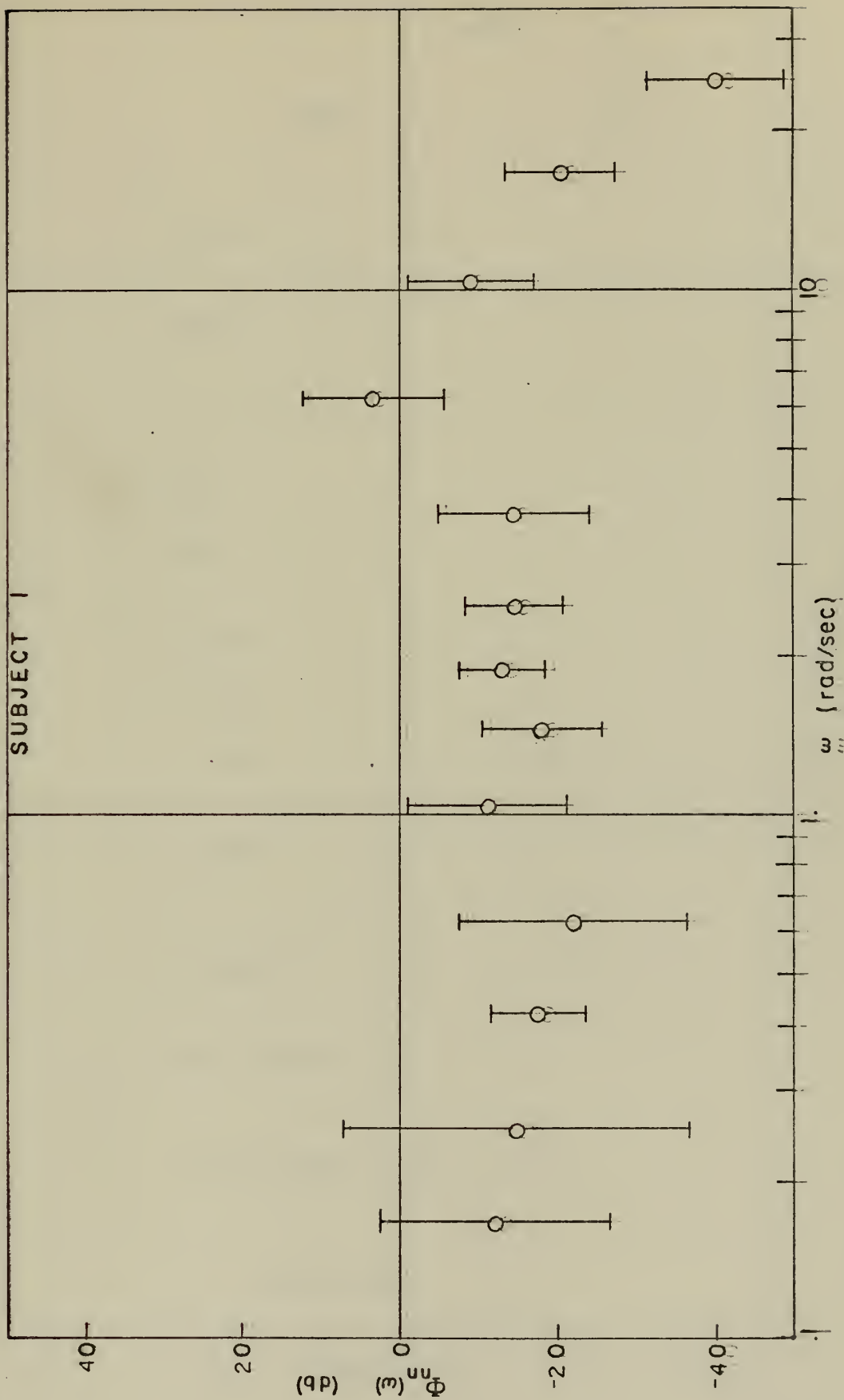


FIGURE 15



AMPLITUDE of HUMAN DESCRIBING FUNCTION for  $Y_C(s) = \frac{1}{s}$

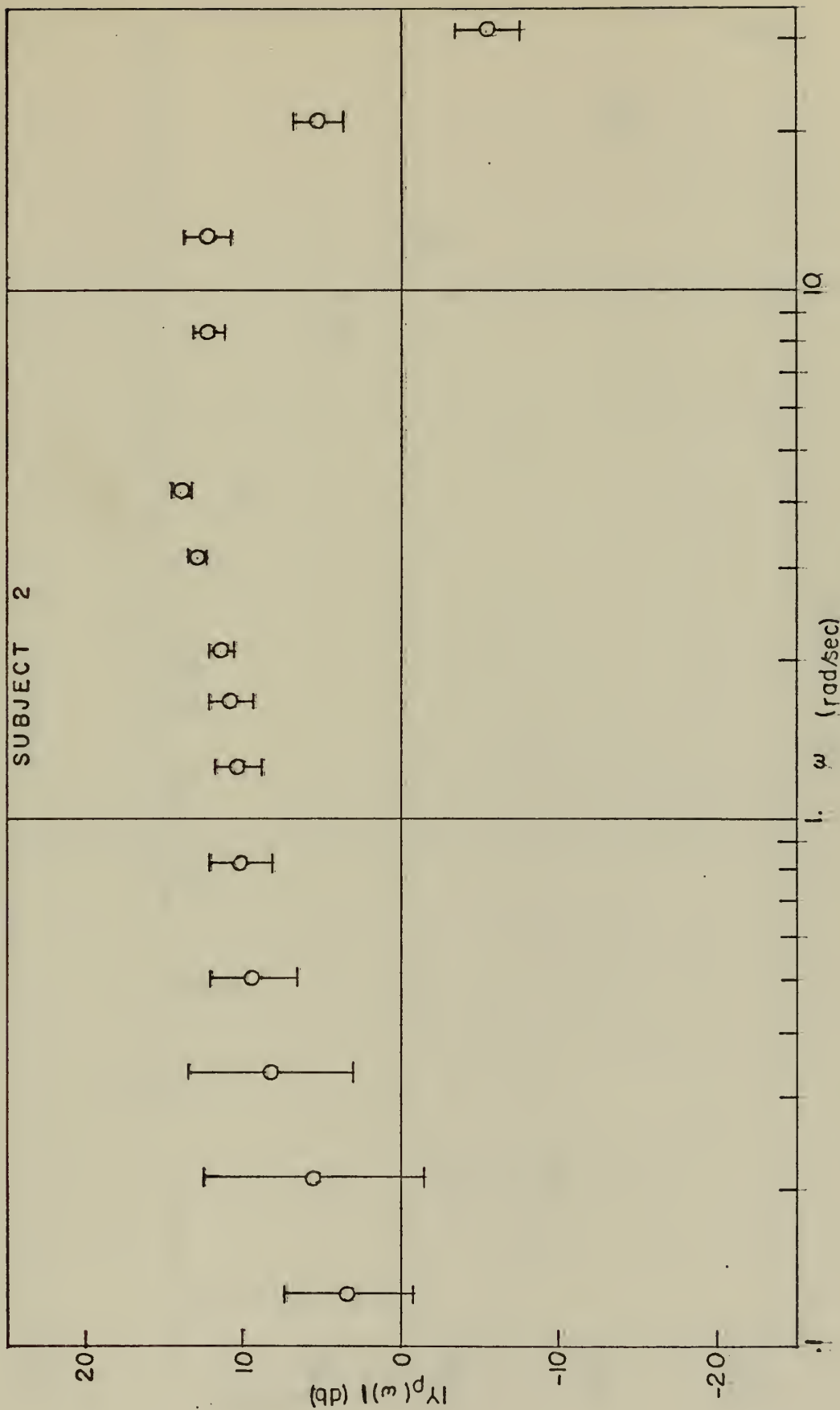
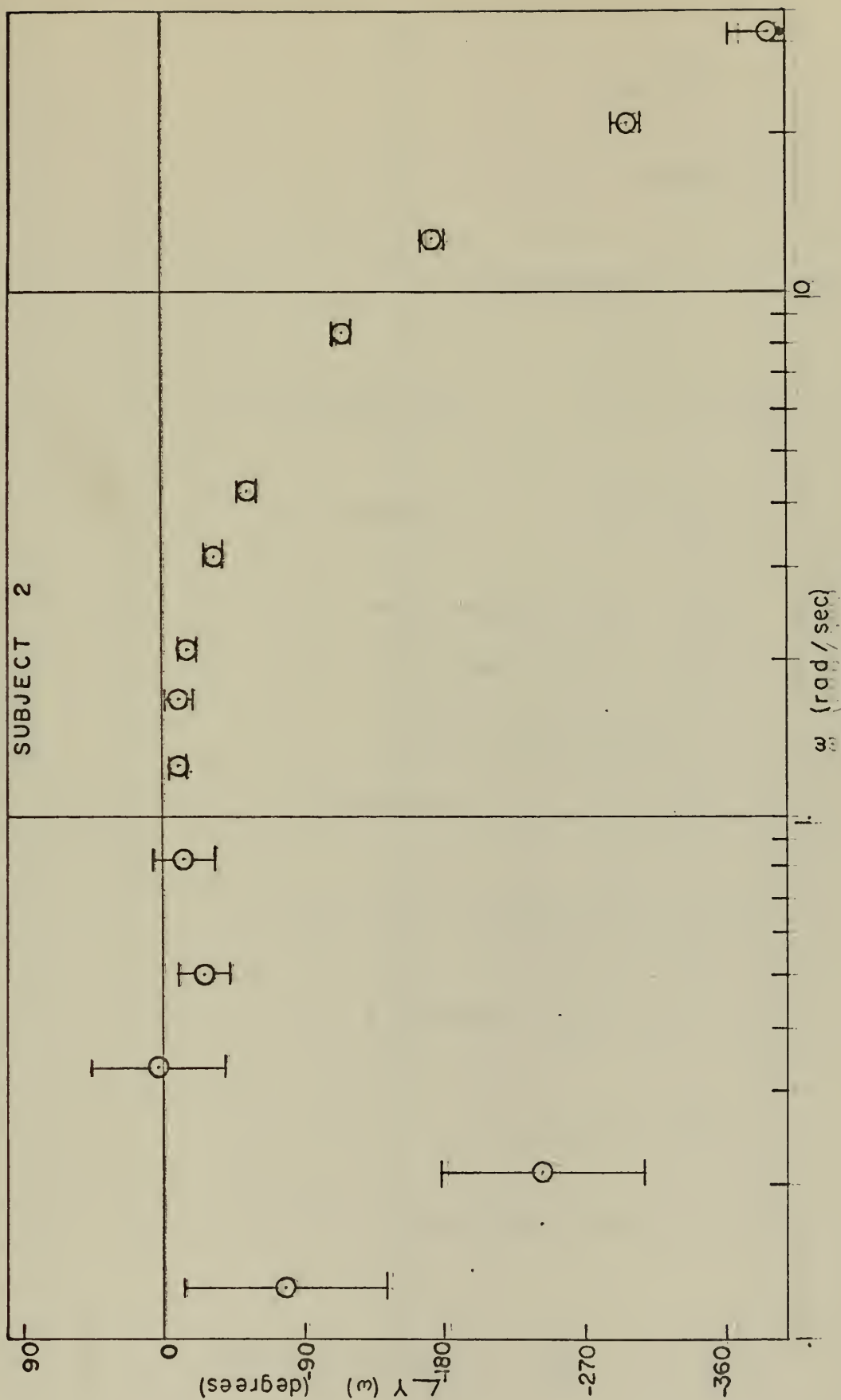


FIGURE 16



PHASE of HUMAN DESCRIBING FUNCTION for  $Y_C(s) = \frac{1}{s}$



$\omega$  (rad/sec)  
FIGURE 17





REMNPANT of HUMAN DESCRIBING FUNCTION for  $\gamma_c(s) = \frac{1}{s}$

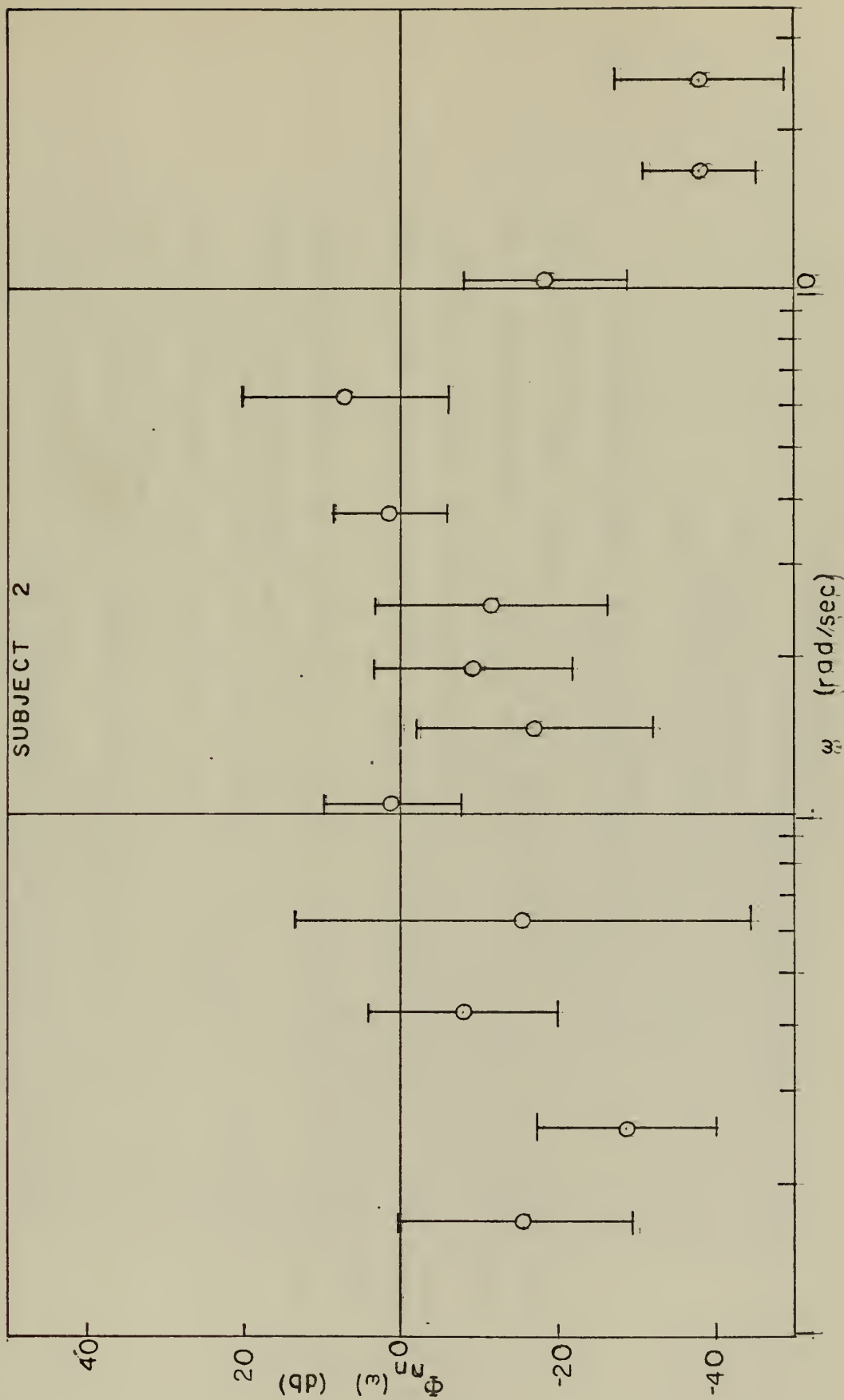


FIGURE 18



```

100 F0RMA1 (9X,1F10.5,1F15.7)
101 F0RMA1 (1H1,8X,1HK,10X,6HAMP(K),12X,4HW(K),/(7X,14,6X,1F10.5,4X,1
1F15.7))
102 F0RMA1 (1H1,4X,1HK,10X,5HAE(K),10X,5HBF(K),10X,5HA1(K),10X,5HB1(K)
1,10X,5HA2(K),10X,5HBP(K),/(15,2X,6F15.7))
103 F0RMA1 (///,4X,1HK,11X,4HW(K),10X,5HYP(K),8X,7HAPH(K),10X,5HYC(K)
1,8X,7HAPHAC(K),/(15,5F15.7))
104 F0RMA1 (///,4X,6HPUTSQ=,1F15.7,4X,6HERRSQ=,1F15.7,4X,6HONESQ=,1F15.
17,4X,6HONESY=,1F15.7,/(29X,6HERRSY=,1F15.7,4X,6HONESY=,1F15.7,4X,6
2HT,9SY=,1F15.7)
105 F0RMA1 (///,4X,1HJ,11X,4HW(J),7X,8HPHIPP(J),8X,8HPHINN(J),9X,6HYP
1C(J),/(15,4F15.7))
106 F0RMA1 (1H1,30X,394DATA LISTED BELOW IS A MEASURE OF POWER,/,
133X,26HNORMALIZED TH (VOLT)(VOLT),/(33X,24HPHASE DATA IS IN DEGREE
25)
107 F0RMA1 (1H1,30X,31HDATA FOR YP, YC, PHIPP IS IN DB,/,33X,24HPHASE
1DATA IS IN DEGREES)
108 F0RMA1 (6X,12)
109 F0RMA1 (///,30X,69HGRAPH OF PILOT DESCRIBING FUNCTION VERSUS FREQ
1UENCY IN RADIANS/SECOND)
110 F0RMA1 (///,33X,62HGRAPH OF CONTROLLED DYNAMICS VERSUS FREQUENCY
1IN RADIANS/SECOND)
111 F0RMA1 (///,27X,75HGRAPH OF PILOT DESCRIBING FUNCTION PHASE VERSU
1S FREQUENCY IN RADIANS/SECOND)
112 F0RMA1 (///,30X,69HGRAPH OF CONTROLLED DYNAMICS PHASE VERSUS FREQ
1UENCY IN RADIANS/SECOND)
113 F0RMA1 (///,30X,39HGRAPH OF REMNANT POWER VERSUS FREQUENCY)
COMMON KOUT,KUIT,C1,C1,CONST,DELT,I,PI
COMMON PUT,PUTR,E,ENE,IAB
COMMON PUTSQ,ERRSQ,ONESQ,IABSQ,ERRSY,ONESY,TW9SY
COMMON AMP(29),A(29),A(29),B(29),C(29),D(29)
COMMON AF(29),FE(29),A1(29),B1(29),A2(29),B2(29)
COMMON FE(29),FP(29),FC(29)
COMMON YP(29),PHAP(29),YC(29),PHAC(29)
COMMON AYP(15),BYP(15),AYC(15),BYC(15),YPC(15)

```



```

COMMON PHIDP(15),PHINV(15)
COMMON NREMW,AMPLI
COMMON U(14,5),JXY(2),U1(13,2)
CALL DISABE
CALL SETPBT (4HPC00,.0555,4HP001,.0555,4HP003,.0555,4HP005,.0555)
CALL SETPBT (4HPC30,.2000,4HP031,.2000,4HP033,.1000)
CALL SETPBT (4HPC17,.1600,4HP002,.5000,4HP024,.7500)
LDA 4+3
STA 052
BRJ 45
RRM 4+1
HLT
CALL INTP
LDA KUIT
BRX 4-4,0
4 CALL RESET(1000)
DELT = 0.02
READ IF FREQUENCY OTHER THAN INPUT FOR REMNANT SCALING
READ(5,108) NREMW
FIRST 14 FREQS, GENERATE THE INPUT
AMPLITUDE OF LAST 15 FREQS, HAVE VALUE 0.0
READ(5,100) (AMP(K),W(K),K=1,29)
WRITE(6,101)(K,AMP(K),W(K),K=1,29)
TIME SET 52 THAT PROBLEM IS IN STEADY STATE INITIALLY
1 9,SET = -1080.
SCALES INPUT 50 THAT NR VALUE GREATER THAN + 03 = 0.999
CONST = 0.125
SET INITIAL VALUES FOR INPT, TIME AND ZERO ARRAYS
PUT = 0.0
PUTR = 0.0
CALL DAC(1,PUT,2,PUTR)
T = 150.
PI = 3.1415927
AMPLI SCALES THE REMNANT AMPLITUDE
AMPLI = 0.10

```



```

*
*
DETERMINE ALL FREQS. SO CAN LATER DETERMINE MAGNITUDES OF
  OUTPUTS AT ALL FREQUENCIES
DO 10 K=1,29
  A(K) = SIN(W(K)*DELT)
  P(K) = SORT(1.-A(K)**2)
  C(K) = SIN(W(K)*ENSET)
  D(K) = SORT(1.-C(K)**2)
  PUT = PUT + AXP(K)*C(K)
10 CONTINUE
*
  GIVES INITIAL VALUE TO DAC
  PUT = PUT*CONST
  PUTR = AMPLI*(NREMA)
  E DENOTES OUTPUT FOR EN
  1 DENOTES OUTPUT FOR PN
  2 DENOTES OUTPUT FOR CN
  INITIALIZE ALL STORAGE AREAS
DO 20 K=1,29
  AE(K) = 0.
  A1(K) = 0.
  A2(K) = 0.
  BE(K) = 0.
  P1(K) = 0.
  P2(K) = 0.
20 CONTINUE
  FOUNT = 0
  KUIT = 0
  PUTSC = 0.
  ERRSQ = 0.
  GNESS = 0.
  TMPSC = 0.
  ENABLE PROBLEM TO BE STARTED WHEN OPERATOR READY
  OUTPUT(101) ISET DSO SWITCH TO START RUN!
511 IFLG1 = TEST(3)
   IF (IFLG1.GT.0) GO TO 511
   CALL COMPUTE

```





```

CALL ENABLE
30 CONTINUE
IF (KUT-EG.1) GO TO 31
GO TO 30
31 CONTINUE
CALL DISABLE
CALL HOLD
CALL ACK(4,PUTSQ,5,ERRSQ,6,8NESQ,7,TWOSQ)
CALL RESET(1000)
WRITE(6,102) (K,AE(K),BE(K),A1(K),B1(K),A2(K),B2(K),K=1,29)
ERRSQ = ERRSQ*4.
ERRSY = ERRSQ/PUTSQ
ONESY = ONESQ/PUTSQ
TWOSSY = TWOSQ/PUTSQ
RMSCON SCALES THE ROOT MEAN SQUARE VALUE
RMSCON = 1000.
PUTSQ = PUTSQ*RMSCON
ERRSQ = ERRSQ*RMSCON
ONESQ = ONESQ*RMSCON
TWOSSQ = TWOSSQ*RMSCON
FE IS NOTATION FOR EN
FP IS NOTATION FOR PN
FC IS NOTATION FOR CN
DO 40 K=1,29
FE(K) = AE(K)**2 + BE(K)**2
FP(K) = A1(K)**2 + B1(K)**2
FC(K) = A2(K)**2 + B2(K)**2
YF(K) = SQRT(FP(K)/FE(K))
YC(K) = SQRT(FC(K)/FP(K))
PHAP(K) = 57.3*(ATAN(BE(K),AE(K)) - ATAN(B1(K),A1(K)))
PHAC(K) = 57.3*(ATAN(B1(K),A1(K)) - ATAN(B2(K),A2(K)))
* ASSUME PHASE LEAD LESS THAN 180 DEGREES TO CORRECT FOR LOSS OF
* PHASE INFORMATION IN ARC-TANGENT ROUTINE
IF (PHAP(K)-LT.180.) GO TO 42
PHAP(K) = PHAP(K) - 360.

```



```

42 IF (PHAC(K).LT.180.) GO TO 44
   PHAC(K) = PHAC(K) - 360.
44 IF (PHAP(K).GT.-180.) GO TO 41
   PHAP(K) = 360. + PHAP(K)
41 IF (PHAC(K).GT.-180.) GO TO 43
   PHAC(K) = 360. + PHAC(K)
43 CONTINUE
40 CONTINUE
   CONS2 = 0.046757339
   *
   DETERMINE REAL AND IMAGINARY PORTIONS OF YP AND YC
   DO 45 K=1,14
     AYP(K) = YP(K)*COS(PHAP(K)/57.3)
     BYP(K) = YP(K)*SIN(PHAP(K)/57.3)
     AYC(K) = YC(K)*COS(PHAC(K)/57.3)
     BYC(K) = YC(K)*SIN(PHAC(K)/57.3)
45 CONTINUE
   POLATE IS FACTOR USED FOR INTERPOLATION OF REMNANT SIGNAL
   DO 46 K = 1,13
     J = K + 15
     POLATE = (W(J) - W(K))/(W(K+1) - W(K))
     YPC(K) = (1. + ((AYP(K) + AYP(K+1))*POLATE) * ((AYC(K) + AYC(K+1))
1) *POLATE)) - (((BYP(K) + BYP(K+1))*POLATE) * ((BYC(K) + BYC(K+1))*
2POLATE)))*2 + (((BYP(K) + BYP(K+1))*POLATE) * ((AYC(K) + AYC(K
3+1))*POLATE)) + (((AYP(K) + AYP(K+1))*POLATE) * ((BYC(K) + BYC(K+1
4))*POLATE)))*2
     PHIPP(K) = FP(J)/(2.*T)
     PHINN(K) = PHIPP(K)*YPC(K)*CONS2
46 CONTINUE
   WRITE (6,106)
   *
   WRITE (6,103) (K,W(K),YP(K),PHAP(K),YC(K),PHAC(K),K=1,14)
   *
   WRITE (6,105) (J,W(J+15),PHIPP(J),PHINN(J),YPC(J),J=1,13)
   *
   WRITE (6,104) PUTSQ,ERRSQ,ONESQ,TWOSQ,ERRSY,ONESY,TWOSY
   DO 50 K=1,14
     U(K,1) = ALG010(W(K))
     U(K,2) = 20.*ALG010(YP(K))

```







```

SUPROUTINE INTR
COMMON KOUNT,KUIT,C1,D1,C9NS1,DELT,T,PI
COMMON PUT,PUTR,E,ONE,TWO
COMMON PUTSQ,ERRSQ,ONESQ,TWOSQ,ERRSY,ONESY,TWOSY
COMMON AYP(29),W(29),A(29),B(29),C(29),D(29)
COMMON AE(29),BE(29),A1(29),B1(29),A2(29),B2(29)
COMMON FF(29),FP(29),FC(29)
COMMON YP(29),PHAP(29),YC(29),PHAC(29)
COMMON AYP(15),BYP(15),AYC(15),BYC(15),YPC(15)
COMMON PHIPP(15),PHINN(15)
COMMON ARETN,AMPLI
KOUNT = KOUNT+1
* FIRST 30 SECONDS IS WARM-UP TIME
IF (KOUNT.GT.9000) GO TO 30
* NEXT 150 SECONDS IS RUN TIME
IF (KOUNT.GT.1500) GO TO 20
CALL DAC(1,PUT,2,PUTR)
* FROM THIS POINT TO JG8 TO 32J IS WARM-JP PARTIEN
PUT = C.0
DO 10 K=1,29
C1 = A(K)*D(K) + B(K)*C(K)
D1 = B(K)*D(K) - A(K)*C(K)
C(K) = C1
D(K) = D1
PUT = PUT + AYP(K)*C1
10 CONTINUE
PUT = PUT*C9NS1
PUTR = AMPLI*C(NREMW)
GO TO 32
* THIS PARTIEN IS RUN PARTIEN
20. CALL ADDA(E,3,PUT,2)
PUT = C.0
DO 21 K=1,29
C1 = A(K)*D(K) + B(K)*C(K)
D1 = B(K)*D(K) - A(K)*C(K)

```





```

C(K) = C1
D(K) = D1
IF (K.LE.14) PUT=PUT+AMP(K)*C1
AE(K) = AE(K) + E*D1
A1(K) = A1(K) + ONE*D1
A2(K) = A2(K) + TWO*D1
PE(K) = PE(K) + E*C1
P1(K) = P1(K) + ONE*C1
P2(K) = P2(K) + TWO*C1
21 CONTINUE
PUT = PUT*CONS1
PUTR = AMPLI*C(NRENW)
GO TO 32
30 KUIT SETS PROGRAM END TIME--TAKES IT OUT OF INTERRUPT CONTROL
32 KUIT = 1
32 RETURN
END

```

The following subroutine calls analog-to-digital conversion and vice versa to conserve computer time between interrupts:



X1	EOU	1	
X2	EOU	2	
A	EOU	5	
B	EOU	4	
RADDA	PZE		SSETUPN
	3BY	4	
AD	PZE		
NAD	PZE		
DA	PZE		
NNA	PZE		
	SKR		FINFLO
	BRU		\$-1
	LDA		BEFIN
	STA		O+O
	STA		C41
	LCA		*RAD
	STA		CPUNT
	SKR		COUNT
	BRU		+2
	BRU		DTCA
	LLSA		15
	ADD		=ADCEM
	STA		APCA
	SKN		FINFLO
	BRU		\$-1
	STZ		FINFLG
	ENH		034001
	PST		APCA
	LDX		=00100000,1
	SKN		FINFLG
	BRU		\$-1
ITER	LCA		ADPUF,1
	COPY		(0,3)
	FLA		0.0



STD	*AD
MPT	AD
SKR	COUNT
BRX	ITER,1
LDA	*ADA
ARG	=077700000
COPY	(A,X1)
COPY	(C,A)
STA	DACM,1
BRX	#+2,1
RRR	ADDA
STZ	DCH
BRJ	#+2
MPT	DA
MPB	DCH
LOP	*CA
CRSD	9
ARGD	15
COPY	(-A,A)
COPY	(A,X2)
COPY	(E,A)
ARSA	C,2
ETR	=077777000
ADD	DCH
STA	DACM,1
BRX	RTPL,1
STZ	FINFLG
LDA	*ADA
LLSA	15
ADD	=DACM
STA	DACA
EDM	035001
RET	DACA
GRR	ADDA
SON	FIN









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## DOCUMENT CONTROL DATA - R &amp; D

(Security classification of title, body of abstract and indexing annotation must be entered when the overall report is classified)

1. ORIGINATING ACTIVITY (Corporate author)

Naval Postgraduate School

Monterey, California 93940

2a. REPORT SECURITY CLASSIFICATION

Unclassified

2b. GROUP

3. REPORT TITLE

A Hybrid Computer Technique for Measuring Human Describing Functions and Remnant in Closed-Loop Tracking Tasks

4. DESCRIPTIVE NOTES (Type of report and, inclusive dates)

Engineer's Thesis; June 1972

5. AUTHOR(S) (First name, middle initial, last name)

Roy D. Warren

6. REPORT DATE

June 1972

7a. TOTAL NO. OF PAGES

7b. NO. OF REFS

8a. CONTRACT OR GRANT NO.

9a. ORIGINATOR'S REPORT NUMBER(S)

b. PROJECT NO.

c.

9b. OTHER REPORT NO(S) (Any other numbers that may be assigned this report)

d.

10. DISTRIBUTION STATEMENT

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11. SUPPLEMENTARY NOTES

12. SPONSORING MILITARY ACTIVITY

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13. ABSTRACT

The measurement of the human describing function and remnant in a compensatory tracking task is undertaken. These measurements are obtained through the application of the fast Fourier transform technique on a hybrid (analog-digital) computer. This method processes the data in real time with minimal core storage and the results are available immediately upon completion of the tracking run.



KEY WORDS	LINK A		LINK B		LINK C	
	ROLE	WT	ROLE	WT	ROLE	WT
1. Human Describing Function						
2. Human Response Function						
3. Human Transfer Function						
4. Pilot Response						
5. Describing Function						
6. Human Remnant						
7. Remnant						





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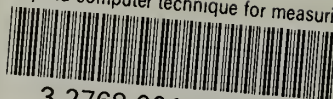
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